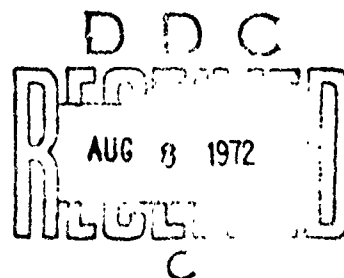




FRANK J. SEILER RESEARCH LABORATORY

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AD 746922

AN ENGINEER'S GUIDE

TO BUILDING NONLINEAR FILTERS

VOLUME II

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Calvin Hecht
Capt Kenneth D Senne

SEE
AD 746921

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PROJECT 7904

AIR FORCE SYSTEMS COMMAND
UNITED STATES AIR FORCE

230

Additional Appendix A.

A Two-Dimensional Point-Mass Program for the Passive Receiver Problem

This program was written in a format which is easily adaptable to an arbitrary two-dimensional problem. The program is set up to produce data for the movie of Chapter VI. Contour maps are printed to provide information regarding the density shapes and data is written on magnetic tape.

The program was compiled on one of the Kirtland AFB CDC 6600's using the RUN compiler. The random number generator used is similar to but inferior to the one described in Chapter IV.

Programmer: K.D. Senne


```

0000113 3 114K4AT(N/N-2),44,124SIGV4(N/N-2),5X,74SQ(L4W,3X,34PHI //)
0000116 IF (.NOT.P5) CALL OUTPUT(N)
0000155 IF (.NOT.P5) WRITE (PR,60(50) IN, XL(1), PL(1), PL(2), XL(1),
1 PL(1), PL(2), XL(2), PL(3), XL(2), PL(2), PL(3)
6005 FOR4AT(14H LINEARIZED CHECK-14,16H ITERATION(S)... ,7X,6E10.3/
1 45X,6E11.3/)
C CALL TIMER(1)
0000155 DO 400 N=NS,NMAX
0000157 IF (.NOT.MLF) CALL S4LFT(N,EJ,JC)
0000163 IPRTS=5
0000164 IF (MLF) GO TO 310
0000166 CALL FILL4IN,Z,R,JC,H1,H2,H3)
0000170 CALL ELLPSE(N,IY1,IX2,Z,Y4,PN,NPTS,XKN)
0000176 IF (SCANX) GO TO 200
0000206 CALL YCNV(MC,N,H,EJ,JC,H1,H2,H3,IX1,IX2)
0000210 GO TO 250
0000221 201 CALL KCNV(MC,N,H,EJ,JC,H1,H2,H3,IX1,IX2)
0000222 25 IF (C-1.1E-20) GO TO 300
0000234 WRITE (PR,50130) '1, Z(1), Z(2)
0000240 60103 FOR4AT(17X,6HAT N =,I4,7H, Z = (,E10.3,1H,E10.3,154) HAS DISCARD
0000251 1.)
0000251 IPRTS=NPTS-1
0000253 IF (NPTS.GT.0) GO TO 100
0000255 STOP
0000256 301 CALL SCALE(M,EJ)
0000260 311 CALL STATE(N)
0000262 IF (.NOT.IS) CALL WRITE(MC,N,H,EJ,IX1,IX2)
0000270 IF (.NOT.S-OR-MLF) CALL FLIN(XL,PL,XL1,PL1,N,IN)
0000277 IF (.NOT.MLF) GO TO 32.
0000301 YN(1)=XL(1)
0000302 N(2)=XL(2)
0000304 321 IF (.NOT.(MLF.AN).CP) GO TO 330
0000307 DETC=PL(1)*PL(3)-PL(2)*PL(2)
0000312 CA=1./((HQP+S121(1E10))
0000317 CALL YCNV(PL,PL1,2)
0000321 K=1
0000322 DO 325 J=1,M4X
0000324 JP(2)=JC(J)
0000326 DO 325 I=1,M4X
0000327 XP(1)=JC(I)
0000331 CALL XUNP4MIXP,XUN1,RL4YN1,YN1,XT)
0000334 LG(K)=CA*NBP*AL(XT,XL,PL1)
0000342 K=K+1
0000347 CALL WK4M1(J,J,J,E,M4X,13)
0000351 331 IF (CP.AND.(NOT.MLF)) CALL WK4PLT(JC,JC,EJ,M4M4X,13)
0000361 IF (PS) GO TO 350
0000363 CALL WUPJ(N)
0000364 WRITE (PR,50350) I4, XL(1), PL(1), PL(2), XL(1), PL1(2),
1 XL(2), PL(2), PL(3), XL(2), PL1(2), PL1(3)
0000422 351 CALL WC4V(MC,I,X,YN,E4C,V4C)
0000426 401 CONTINUE
0000431 451 Y=0
0000432 IF (.NOT.IS) CALL MCRTIE(MC)
0000435 IF (MLF.AN).(400(MC,MP1).EQ..02,MC.EQ.NMC)) WRITE (PR,60150) 40
0000455 FOR4AT(14H,25X,22H4MONTE CARLO AVERAGE OF ,15,
1 44H SAMPLE PATH(S) OF THE LINEARIZED PREDICTOR
2 1723X,14H,213X,134CUM AVG ERROR,17.6,194CUM AVG CORRELATION //)

```



```

000455 IF (.NOT.MLF) WRITE (PR,50200) MC
000464 6020 FORMAT(1H1,39X,22HMONTE CARLO AVERAGE OF ,15,16H SAMPLE PATH(S),
1 //25X,14H,23X,13HCOM AVG ERROR ,17X,19HCOM AVG CORRELATION //)
000464 IF (MLF) AN7.(MON7MC,MP0).E7.0.DR.MC7EQ.N7MC) GO TO 475
000476 IF (MLF) GO TO 550
000477 475 CONTINUE
000477 I=1
000500 NM3=3*HMAX
000500 10 500 J=1,NM3,3
000502 <=I+1)/2
000503 WRITE (PR,50300) K, EMC(I), VMC(J), VMC(J+1), EMC(I+1), VMC(J+1),
000505 1 VMC(J+2)
000527 6030. FORMAT(21X,110,15X,E15.8,10X,2E15.8/+6X,E15.8,10X,2E15.8/)
000527 50 I=I+2
000533 550 CONTINUE
000533 C CALL TIMER(3)
000533 60. CONTINUE
000536 IF (.NOT.TS) REMIND TP
000541 CALL EXIT
000542 EN

```

```

26MAR71 10
KFNLF 86
KFNLF 87
26MAR71 19
26MAR71 20
26MAR71 21
26MAR71 22
KFNLF 88
KFNLF 89
KFNLF 90
KFNLF 91
KFNLF 92
KFNLF 93
KFNLF 94
KFNLF 95
26MAR71 22
KFNLF 96
KFNLF 97
KFNLF 98
KFNLF 99
KFNLF 100

```

PROGRAM LENGTH INCLUDING I/O BUFFERS
022701

FUNCTION ASSIGNMENTS

```

ST. TIME IT ASSIGNMENTS
2 - 00165 1.0 - 000227 250 - 000235
- 00261 320 - 000354
- 00043 450 - 000534
6 - 00057 600.0 - 000612
60.0 - 00063 6.20 - 000702

```

BLD. NAMES AND LENGTHS

```

C41 - 00015 CM2 - 00017 CM4 - 00016
C45 - 00020 CM6 - 00017
C49 - 00076 CM10 - 00017

```

VARIABLE ASSIGNMENTS

```

A - 00010013 C
DETR - 014005 E6
FL - 014005 H1
H3 - 014005 I
IX2 - 014005 IZ
K - 014005 MC
HMAX - 014005 MP0
NMC - 014005 N3
NPTS - 014005 NS
PLI - 014005 PL1
PE - 014005 PL2
QE - 014005 PL3
PUN1 - 014005 PL4
SQRT - 014005 PL5
TWOP1 - 014005 PL6
XKH - 014005 PL7

```

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```

XLM - 000.13C10 XLS - 070007C10 XL1 - 012517 XH - 000001C07
XP - 014563 XT - 014565 XZ - 000000C12 YN - 000123C02
FN1 - 000.1-C32 FN2 - 010103C 2 Z - 000102C04

START OF CONSTANTS
000545

START OF TEMPORARIES
000711

START OF INDIRECTS
000721

UNUSED COMPILER SPACE
014000

```



```

000250      U3=(X0-X(I-1))*(X0-X(I))*(X0-X(I+1))/((X(I+1)-X(I-1))*(X(I+1)-X(I)
1) *X(I+1)-X(I+2)))
000263      U4=(X0-X(I-1))*(X0-X(I))*(X0-X(I+1))/((X(I+2)-X(I-1))*(X(I+2)-X(I)
1) *X(I+2)-X(I+1)))
C
C      COMPUTE Y-ARRAY LAGRANGIAN COEFFICIENTS.
C
000276      V1=(Y0-Y(J))*(Y0-Y(J+1))*(Y0-Y(J+2))/((Y(J-1)-Y(J))*(Y(J-1)-Y(J+1)
1) *Y(J-1)-Y(J+2)))
000312      V2=(Y0-Y(J-1))*(Y0-Y(J+1))*(Y0-Y(J+2))/((Y(J)-Y(J-1))*(Y(J)-Y(J+1)
1) *Y(J)-Y(J+2)))
000325      V3=(Y0-Y(J-1))*(Y0-Y(J))*(Y0-Y(J+2))/((Y(J+1)-Y(J-1))*(Y(J+1)-Y(J)
1) *Y(J+1)-Y(J+2)))
000340      V4=(Y0-Y(J-1))*(Y0-Y(J))*(Y0-Y(J+2))/((Y(J+2)-Y(J-1))*(Y(J+2)-Y(J)
1) *Y(J+2)-Y(J+1)))
C
C      INTERPOLATE Z VALUE AS MEAN OF X AND Y INTERPOLATIONS.
C
000353      ALAG30=0.5*(Z(I-1,J-1)*U1+Z(I,J-1)*U2+Z(I+1,J-1)*U3+Z(I+2,J-1)*U4
1) *V1+Z(I-1,J)*U1+Z(I,J)*U2+Z(I+1,J)*U3+Z(I+2,J)*U4)*V2+Z(I-1,J+1)
2) *U1+Z(I,J+1)*U2+Z(I+1,J+1)*U3+Z(I+2,J+1)*U4)*V3+Z(I-1,J+1)*U1+Z(I
3) *U2+Z(I+1,J+2)*U3+Z(I+2,J+2)*U4)*V4+Z(I-1,J+1)*U1+Z(I,J+1)*U2+Z(I+1,J+1)
4) *U3+Z(I+2,J+1)*U4)*V5+Z(I-1,J+1)*U1+Z(I,J+1)*U2+Z(I+1,J+1)*U3+Z(I+2,J+1)
5) *U4+Z(I+2,J+1)*U5)*V6+Z(I-1,J+1)*U1+Z(I,J+1)*U2+Z(I+1,J+1)*U3+Z(I+2,J+1)
6) *U4+Z(I+2,J+1)*U5)*V7+Z(I-1,J+1)*U1+Z(I,J+1)*U2+Z(I+1,J+1)*U3+Z(I+2,J+1)
7) *U4)
000474      RETURN
C
000474      WRITE (6,13) 'X,Y
000504      ALAG30=0.0
000505      PAUSE
000513      RETURN
C
000515      15      FORMAT (974 ***** FUNCTION ALAG30.  INCORRECT VALUE OF NX AND/OF
1) Y.  PAUSE WITH ALAG30 SET TO 0.0.  *****/64 NX = I4,3H,  N/ = ,
2) I4,1H,1) X)
000515      END

```

SUBPROGRAM LENGTH

001053

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

1	-	00012	2	-	00017	3	-	00023	4	-	00026
5	-	00037	6	-	00033	7	-	00035	8	-	00042
9	-	00046	10	-	00051	11	-	00053	12	-	00056
13	-	00067	14	-	000475	15	-	000524			

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS

ALAG30	-	00136	I	-	00137	J	-	00136	L	-	001641
M	-	00142	NY	-	00080	U1	-	00143	U2	-	001644
U3	-	00145	U4	-	00146	V1	-	00147	V2	-	001653
V3	-	00151	V4	-	00152						

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KFNL F
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 KFNL F

SUBROUTINE AX(S,A,M)
 DIMENSION S(1), A(1), T(4)
 COMMON / CH1/ NT(8), NP
 COMMON / CH3/ RT(6), Q(3)
 T(1)=S(1)
 T(2)=S(2)
 IF (NP.EQ.1) GO TO 19
 T(2)=S(3)
 T(3)=S(2)
 T(4)=S(4)
 1. CALL FTAF(T,Q,A,NP,2)
 RETURN
 END

SUBPROGRAM LENGTH
 00043

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS
 10 - 00017

BLOCK NAMES AND LENGTHS
 CH1 - 00011 CM3 - 00011

VARIABLE ASSIGNMENTS
 NP - 00013C01 NT
 T - 00037

START OF CONSTANTS
 00026

START OF TEMPORARIES
 00030

START OF INDIRECTS
 00032

UNUSED COMPILER SPACE
 037100

- 000006C02 RT - 000100C02

KFNLF 114
 KFNLF 115
 KFNLF 116
 KFNLF 117

000005 FUNCTION CUMAVG(AV,X,I)
 000011 CUMAVG=((I-1.)*AV+X)/I
 000012 RETURN
 000012 END

SUBPROGRAM LENGTH
 000026

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS
 **HVG - 000025

START OF CONSTANTS
 000015

START OF TEMPORARIES
 000017

START OF INDIRECTS
 010025

UNUSED COMPILER SPACE
 037200

```

000004 SUBROUTINE DENSITY(JC,EJ)
000004 REAL JC(1), EJ(1), CI(1), C(1), X(2), X(1(2), C(3)
000004 ***** DENSITY
000004 REAL NORMAL
000004 COMMON / CM6/ PI, TWOPI
000004 COMMON / CM1/ M, MSD, MMAX
000004 COMMON / CM2/ Y(9), YN1(2), QUN_(4)
000004 COMMON / CM7/ M4, X4(2), PC(30)
000004 M=MMAX*MMAX
000004 Y3=NH*3
000004 DO 10 I=1, 4M
000004 10 EJ(I)=0.
000004 T=0.
000004 K1=1
000004 DO 30 K=1, N3, 3
000004 C(1)=PC(K)
000004 C(2)=PC(K+1)
000004 C(3)=PC(K+2)
000004 CALL MINV(C, CI, 2)
000004 X(1(1))=X(1(1))
000004 X(1(2))=X(1(2))
000004 DETC=C(1)*C(3)-C(2)*C(2)
000004 X=1./((TWOPI*SQRT(DETC))
000004 L=1
000004 DO 20 J=1, MMAX
000004 XP(2)=JC(J)
000004 DO 20 I=1, MMAX
000004 XP(1)=JC(I)
000004 CALL XUP(XP, RUN1, RLAMN1, /N1, X)
000004 Z=GX*NORMAL(X, X4, CI)
000004 EJ(L)=EJ(L)+Z
000004 L=L+1
000004 20 K1=K1+2
000004 30 J=(T*MSD*NSD*RLAMN1(1)*RLA'N1(2))/FLOAT(M*M)
000004 DO 40 I=1, 4M
000004 40 EJ(I)=EJ(I)/D
000004 RETURN
000004 END

```

SUBPROGRAM LENGTH
000175

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLOCK NAMES AND LENGTHS

CH6	-	000102	C41	-	000203	CM2	-	000220	CM7	-	000063
VARIABLE ASSIGNMENTS											
C	-	000155	CI	-	000144	CX	-	000172	0	-	000174
DETC	-	000167	I	-	000163	J	-	000172	<	-	000166
K1	-	000165	L	-	000171	M	-	000171	CM2	-	000161
MMAX	-	000132	CM	-	000171	CM2	-	000163	NSD	-	000001002
N3	-	000162	PC	-	000250	CM4	-	000250	RLAMN1	-	000114003

118
KFNL F
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KFNL F
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19HAR71
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15HAR71
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KFNL F
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KFNL F
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KFNL F
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KFNL F
126
KFNL F
127
KFNL F
128
KFNL F
129
KFNL F
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KFNL F
131
KFNL F
132
KFNL F
133
KFNL F
134
KFNL F
135
KFNL F
118
19HAR71
119
13HAR71
136
KFNL F
137
KFNL F
138
KFNL F
139
KFNL F
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KFNL F
141
KFNL F
120
19HAR71
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KFNL F
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KFNL F
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KFNL F
146
KFNL F
147
KFNL F
148
KFNL F
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KFNL F
150
KFNL F
151
KFNL F

T - 00016+ TWOPI - 000001C01 X - 000151 X4 - 000001C04
 XM1 - 000153 X0 - 000147 YN1 - 000019C03 Y1 - 000003C03
 Z - 000173

START OF CONSTANTS

000127

START OF TEMPORARIES

000132

START OF INDIRECTS

000142

UNUSED COMPILER SPACE

076500

IT*****VARIABLE GIVEN CONFLICTING TYPES

000305


```

000022 SUBROUTINE ECHECK(M,MC,IX,IY,EJ,JC,H1,H2,H3,IX1,IX2,XT,Y,I1,I2,I3)
000022 REAL EJ(1), JC(1), H1(1), H2(1), H3(1), XT(1), Y(1)
000022 ***** ECHECK
000022 INTEGER IX1(1), IX2(1)
000022 ***** ECHECK
000022 DIMENSION EG(4,1), EH(4,1), EI(4,1), XP(2), X(2)
000022 COMMON / CM1/ M, NSD, HMAX, A11, A12, A22, JPTS, NPQ
000022 COMMON / CM2/ JN2(2), RLAMN2(2), KUN2(4)
000022 COMMON / CM3/ PT(9), CP
000022 COMMON / CM4/ KNI(3)
000022 LOGICAL CP
000022 INTEGER PR
000022 DATA (PR=3)
000022 HMAX=0.
000022 XH=XT(1)
000022 FH=XT(2)
000022 K=1
000022 DO 200 J=1,MHAX
000022 XP(2)=JC(J)
000022 DO 200 I=1,MHAX
000022 XP(1)=JC(I)
000022 CALL XUPRCH(XP,XPJ2,RLAMN2,YN2,X)
000022 Z1=Y(1)+H1(K)
000022 Z2=Y(2)+H2(K)
000022 Z3=Y(3)+H3(K)
000022 Z4=Y(4)+H4(K)
000022 Z5=Y(5)+H5(K)
000022 Z6=Y(6)+H6(K)
000022 Z7=Y(7)+H7(K)
000022 Z8=Y(8)+H8(K)
000022 Z9=Y(9)+H9(K)
000022 IF (O.LT.HMAX) GO TO 1-C
000022 HMAX=0
000022 IM=I
000022 JM=J
000022 EG(K)=Z
000022 Z1=X(1)-XH
000022 Z2=X(2)-YH
000022 J=XH(1)*Z1+Z2*Z3+Z4*Z5+Z6*Z7+Z8*Z9
000022 EH(K)=SQRT(Q)
000022 IF (CP) CALL QMKPLT(J,JC,EJ,HMAX,1J)
000022 MH=HMAX*HMAX
000022 DO 300 I=1,MH
000022 EI(I)=HMAX*EXP(-.5*EH(I)*EH(I))
000022 WRITE (PR,500) MC, M, IX, IY, (EG(I),I=1,MH)
000022 500 FORMAT(9H4TOP MC =,I4,4H N =,I4,4H I =,I4,4H J =,I4,
000022 1 4H, THE KERNEL W(.,.)JN(.,) IS AS FOLLOWS.../(IX,E7.1,14E8.2))
000022 6010 FORMAT(6H4THE THEORETICAL NO. OF STANDARD DEVIATIONS IS AS FOLLO
000022 1 .../(IX,E7.1,14E8.2))
000022 6020 FORMAT(3H4THE ADJUSTED THEORETICAL KERNEL IS COMPUTED TO BE...//
000022 1 (IX,E7.1,14E8.2))
000022 6030 FORMAT(3H4THE MAXIMUM OF W(.,.)JN(.,) IS ,E12.4,
000022 1 2H4AND HAS COORDINATES J =,I4,4H I =,I4,4H J =,I4,4H
000022 6040 FORMAT(4H4THE TRACKING ELLIPSE IS CENTERED AT J =,I4,4H I =,I4,4H
000022 1 5H4 THE RELATIVE ELLIPSE DESCRIPTION IS AS FOLLOWS... /
000022 2 (20X,I5,10X,2I10)
000022 IF (IT.EQ.3) WRITE (PR,61500)
000022 60500 FORMAT(3H4THE SCAN IS ALONG X WITH Y FIXED )

```

000311 IF (ITANE.J) WRITE (02,60600)
 000322 60600 FORNAT(33H)THE SCAN IS ALONG Y WITH X FIXED)
 000322 RETURN
 000323 END

KFNLF
 KFNLF
 KFNLF
 KFNLF

205
 206
 207
 208

SUBPROGRAM LENGTH
 003150

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS
 100 - 000102 60600 - 000332 60120 - 000350 60200 - 000362
 60300 - 000374 60600 - 000410 60500 - 000430 60690 - 000436

ALIASES AND LENGTHS
 C41 - 000111 C42 - 000011 C45 - 000012 C411 - 000003

VARIABLE ASSIGNMENTS

A11 - 000003C01 A12 - 000004C01 A22 - 000005C01 CP - 000011C03
 EG - 000004C01 EH - 000004C01 EI - 000004C01 HMAX - 000011C03
 H1 - 000004C01 H2 - 000004C01 H3 - 000004C01 I - 000011C03
 IM - 000004C01 IY - 000004C01 IX1 - 000004C01 IX2 - 000004C03
 I1 - 000004C01 I2 - 000004C01 J - 000004C01 JM - 000004C03
 K - 000004C01 M4 - 000004C01 NPQ - 000004C01
 PR - 000004C01 PT - 000004C01 RLA4N2 - 000004C02
 RUN2 - 000004C02 X - 000004C02 XH - 000004C02
 XP - 000004C02 XT - 000004C02 Y - 000004C04
 YN2 - 000004C02 Z1 - 000004C02 Z2 - 000004C04

START OF CONSTANTS
 000326

START OF TEMPORARIES
 000444

START OF INDIRECTS
 000452

UNUSED COMPILER SPACE
 035400

*****VARIABLE GIVEN CONFLICTING TYPES
 000023,000023

```

000005      C      SUBROUTINE EIGEN(P,U,XL)
000006      DIMENSION P(1), U(1), XL(1)
000007      DETERMINES EIGENVALUES AND EIGENVECTORS OF P
000008      Q=(P(1)+P(3))/2.
000009      Z=P(1)-P(3)
000010      Z=.5*SQR(3*B+4.*P(2)*P(2))
000011      XL(1)=A+B
000012      XL(2)=A-B
000013      A=P(3)-XL(1)
000014      B=P(2)
000015      C=SQR(A*A+B*B)
000016      IF (C.LT.1.E-40) GO TO 10
000017      U(1)=A/C
000018      U(2)=B/C
000019      GO TO 20
000020      10 U(1)=1.
000021      U(2)=0.
000022      U(3)=0.
000023      U(4)=1.
000024      IF (P(1).GE.P(3)) RETURN
000025      U(1)=0.
000026      U(2)=-1.
000027      U(3)=1.
000028      U(4)=0.
000029      RETURN
000030      20 A=-P(2)
000031      Z=P(1)-P(2)
000032      C=SQR(A*A+B*B)
000033      IF (C.LT.1.E-40) GO TO 30
000034      U(1)=A/C
000035      U(2)=B/C
000036      RETURN
000037      30 U(1)=1.
000038      U(2)=0.
000039      U(3)=0.
000040      U(4)=1.
000041      RETURN
000042      31 U(1)=1.
000043      U(2)=0.
000044      U(3)=0.
000045      U(4)=1.
000046      RETURN
000047      END

```

```

KFNL F 209
KFNL F 210
KFNL F 211
KFNL F 212
KFNL F 213
KFNL F 214
KFNL F 215
KFNL F 216
KFNL F 217
KFNL F 218
KFNL F 219
KFNL F 220
KFNL F 221
KFNL F 222
KFNL F 223
KFNL F 224
KFNL F 225
74AR71 5
74AR71 6
74AR71 7
74AR71 8
74AR71 9
74AR71 10
74AR71 11
74AR71 12
KFNL F 226
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KFNL F 236

```

SUBPROGRAM LENGTH
000154

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS
10 - 000347 29 - 000065 30 - 000107

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS
4 - 000151 8 - 000152 C - 000153

STATE OF CONSTANTS
0.014

STATE OF TEMPORARIES
000122

```

000012      SUBROUTINE ELLIPSE(N,IX1,IX2,Z,Y,P,NPTS,XKN)
000012      DIMENSION IX1(1), IX2(1), P(1), Y(1), Z(1), XKN(1)
000012      C
000012      DETERMINES TRACKING ELLIPSE PARAMETERS
000012      DIMENSION E1(2), XL(2), V1(2), V2(2), ZERO(2)
000012      DIMENSION A(3), AI(3), PI(3), RI(3), XKNIR(3)
000012      DIMENSION DF(4), DH(4), U(4), C1(4), C2(4)
000012      COMMON / CM1/ Y, NSQ, MMAX, AI1, AI2, A22, NZ, YPO, YP, YQ, SCANX
000012      COMMON / CM2/ YN2(2), RLAMN2(2), RUN2(4)
000012      COMMON / CM3/ RI(3)
000012      COMMON / CM6/ TP(2), YSDC
000012      COMMON / C411/ XKN1(3)
000012      COMMON / C44/ XZ(3), XC(2), XK(4)
000012      LOGICAL SCANX
000012      ZERO(1)=0.
000012      ZERO(2)=0.
000012      CALL SX(Y,C1,N)
000012      CALL AX(C1,A,N)
000012      CALL MINV(A,AI,2)
000012      CALL MINV(RI,PI,N)
000012      CALL MINV(PI,2)
000012      CALL JFX(Y,DF,N)
000012      CALL DXF(C,DH,N)
000012      CALL FTA(DF,RI,C1,N7,2)
000012      CALL FTF(Y,AI,C2,C,2)
000012      C JCI 1,3
000012      XN1(1)=C1(1)+C2(1)+PI(1)
000012      XN1(2)=MINV(XKN1,XN1,2)
000012      CALL FITCN(XKN,U,XL)
000012      CALL APPM(1,RIH2,PLA,N2,ZERO,E1)
000012      SCANX=.FALSE.
000012      IF (ABS(E1(2)).GE.ABS(E1(1))) SCANX=.TRUE.
000012      CALL ROTATE(XKN1,XKNIR,RUN2,RLAMN2)
000012      YSDC=NSQC*NSQC
000012      XNS=FLOAT(YSDC)/FLOAT(N*N)
000012      DO 15 I=1,3
000012      XKNIR(I)=XKNIR(I)*XNS
000012      NPTS=0
000012      DO 80 K=1,MMAX
000012      IPQ=K
000012      L=E1(2)*(K-1)/E1(1)+.5
000012      IF (SCANX) L=E1(1)*(K-1)/E1(2)+.5
000012      DO 30 K1=1,MMAX
000012      I=L+1-K1
000012      IF (SCANX) GO TO 23
000012      IF (QUAD(K-1,I,XKNIR).GT.NSQ) GO TO 40
000012      GO TO 30
000012      23 IF (QUAD(I,K-1,XKNIR).GT.NSQ) GO TO 40
000012      30 CONTINUE
000012      40 IX1(K)=I
000012      DO 60 K1=1,MMAX
000012      I=K1+L
000012      IF (SCANX) GO TO 50
000012      IF (QUAD(K-1,I,XKNIR).GT.NSQ) GO TO 70
000012      GO TO 50
000012      50 IF (QUAD(I,K-1,XKNIR).GT.NSQ) GO TO 70
000012      60 CONTINUE
000012      70 IX2(K)=I-IX1(K)
000012      IF (I.GT.0.AND. IX1(K).LE.0) IX2(K)=IX2(K)+1

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000320 NPTS=NPTS+IX2(K)
000323 IF (K.EQ.0) IR=NPTS
000325 IF (SCANX) GO TO 75
000327 IF (QUADTK=1,1,XKNIR).GT.NSQ) GO TO 90
000343 GO TO 90
000343 75 IF (QUAD(L,K-1,XKNIR).GT.NSQ) GO TO 90
000360 8' CONTINUE
000363 91 NPTS=2*NPTS-IR
000365 CALL FX(Y,V1,N)
000370 CALC VMULT(2,2,2,I,OF,Y,V2)
000403 V2(1)=V2(1)-V1(1)
000405 V2(2)=V2(2)-V1(2)
000407 CALL FTA(OF,A1,C1,2,2)
000413 CALL TMULT(XKN,C1,XK,2)
000417 CALL VMULT(2,2,2,1,X,V2,XC)
000420 CALL FTA(OF,RI,C1,NQ,2)
000432 CALL HX(Y,V1,N)
000440 CALL VMULT(NQ,NQ,2,1,OH,Y,V2)
000457 V2(1)=V2(1)-V1(1)+Z(1)
000462 V2(2)=V2(2)-V1(2)+Z(2)
000465 CALL VMULT(2,2,N,NQ,C1,V2,V1)
000475 CALL TMULT(P,Y,V2,1)
000501 V1(1)=V1(1)+V2(1)
000503 V1(2)=V1(2)+V2(2)
000505 CALL TMULT(XKN,V1,V2,1)
000511 XC(1)=XC(1)+V2(1)
000513 XC(2)=XC(2)+V2(2)
000515 RETURN
000516 END

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 KFNLF 323

SUBPROGRAM LENGTH

000622

FUNCTION ASSIGNMENTS

STATEMENT	ASSIGNMENTS					
20	- 000221	37	- 000236	40	- 000241	57
60	- 000302	7	- 000305	75	- 000344	90
90	- 000364					

BLOCK NAMES AND LENGTHS

BLOCK	NAME	LENGTH				
CM1	- 000113	CM2	- 000113	CM3	- 000113	CM6
CM11	- 000133	CM4	- 000136			

VARIABLE ASSIGNMENTS

VARIABLE	ASSIGNMENTS					
A	- 000557	AI	- 000557	C1	- 000503	C2
OF	- 000567	CH	- 000573	E1	- 000536	I
IK	- 000521	K	- 000616	K1	- 000620	L
H	- 000103	HC01	- 000103	NPQ	- 000103	NPTS
NQ	- 000110	NSD	- 000110	NSQ	- 000103	NSQ
PI	- 000556	R	- 000616	PI	- 000551	RLA
RUN2	- 000103	SCANX	- 000103	TP	- 000103	U
V1	- 000542	V2	- 000544	XC	- 000103	XC
XKN	- 000101	XKNI	- 000101	XKNIR	- 000554	XL
XNS	- 000615	XZ	- 000103	YN2	- 000103	ZERO

```

000011      C
SUBROUTINE FILL(N,Z,P,JC,Y1,H2,H3)
REAL Z(1), R(1), JC(1), H1(1), H2(1), H3(1)
GENERATES H TABLES TO REDUCE CONVOLUTION CALCULATIONS
-----
000011      COMMON / CM1/ I(2), YMAX, AT(3), NT(3), NQ
000011      COMMON / CM2/ H2(2), PLANN2(2), RUN2(4)
000011      DIMENSION X(2), XP(2), PI(3), XT(2)
000011      K=1
000012      CALL MINV(R,RI,NQ)
000013      DO 10 J=1,IMAX
000021      XP(2)=JC(J)
000023      DO 10 I=1,IMAX
000024      XP(1)=JC(I)
000026      CALL XUNPR4(XP,RUN2,PLANN2,IN2,X)
000031      CALL FX(X,XT,N)
000044      HI(K)=-XT(1)
000045      H2(K)=-XT(2)
000047      CALL HX(X,XT,N)
000052      X(1)=Z(1)-XT(1)
000057      IF (NQ.GT.1) X(2)=Z(2)-XT(2)
000065      H3(K)=X(1)*RI(1)+X(1)
000067      F=TRQ-GT-1) H3(K)=H3(K)+(2.*X(1)*R(2)+X(2)*R(3))*X(2)
000101      H3(K)=-.5+H3(K)
000104      1) K=K+1
000112      RETURN
000113      END

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 KFNLF 346
 KFNLF 347
 KFNLF 348

SUBPROGRAM LENGTH
 C=145

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLOCK NAMES AND LENGTHS
 CM1 - 000012 CM2 - 000010

VARIABLE ASSIGNMENTS
 AY - 00000001 H3 - 000000 I - 000144 IT - 000000001
 J - 000143 K - 000142 HMAX - 000002001 NQ - 000011001
 NY - 00006001 RI - 000135 PLANN2 - 000002002 RUN2 - 000004002
 X - 000131 XP - 000133 X1 - 000140 YN2 - 000000002

START OF CONSTANTS
 000116

START OF TEMPORARIES
 000121

START OF INDIRECTS
 000125

UNUSED COMPILER SPACE
 036602

*****VARIABLE GIVEN CONFLICTING TYPES

```

000007 SUBROUTINE FINDOP(NM,XM,PC,I,P)
000007 DIMENSION XM(1), PC(1), Y(1), P(1), PINV(3)
000010 NM2=2*NM
000010 DO 10 I=1,2
000013   I(1)=0.
000014   P(1)=0.
000020   P(3)=0.
000021   J=1
000022   DO 20 I=1,NM2,2
000036     Y(1)=Y(1)+PC(J)*XM(I)+PC(J+1)*XM(I+1)
000044     Y(2)=Y(2)+PC(J+1)*XM(I)+PC(J+2)*XM(I+1)
000051     P(1)=P(1)+PC(J)
000054     P(2)=P(2)+PC(J+1)
000057     P(3)=P(3)+PC(J+2)
000061   20 J=J+3
000070   CALL TINV(P,PINV,2)
000103   Y(1)=Y(1)*PINV(1)+Y(2)*PINV(2)
000105   Y(2)=Y(1)*PINV(2)+Y(2)*PINV(3)
000110   Y(1)=Y(1)
000111   RETURN
000112 END

```

SUBPROGRAM LENGTH
000142

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS

I	-	0J0137	J
YT	-	000141	

START OF CONSTANTS

000115

START OF TEMPORARIES

000117

START OF INDIRECTS

000121

UNUSED COMPILER SPACE

036703

PINV - 000133

ANV

— 070136

SYM2

67166

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000010      SUBROUTINE FLIN(XH1,PV1,XL1,PL1,N,IN)
000010      DIMENSION XH1(1), PV1(1), XL1(1), PL1(1)
000010      DIMENSION AK(4), JF(4), OH(4), DH1(4), F(2), H(2), OY(3),
1      TM(4), TH1(4), TH2(4), XH(2)
000010      COMMON / CH1/ NI(8), NP, NQ
000010      COMMON / CH3/ R(3), A(3)
000010      COMMON / CH4/ XI(2), Z(2)
000010      COMMON / CM6/ PI, TWOPI
000010      CALL OHX(XH1,OH,N)
000010      -CALL HX(XH1,H,N)
000020      DO 1000 K=1,NI
000020      IN=K
000025      TM(1)=OH(1)
000026      TM(2)=OH(2)
000027      IF (NO.EQ.1) GO TO 100
000030      TM(2)=OH(3)
000033      TM(3)=OH(2)
000034      TM(4)=OH(4)
000035      10- CALL FTAF(TM,PV1,TH1,2,N)
000042      TM(1)=TM(1)+R(1)
000044      IF (NO.EQ.1) GO TO 200
000052      TM(2)=TM(2)+R(2)
000054      TM(3)=TM(3)+R(3)
000056      20- CALL YINV(TM1,TH2,N)
000061      CALL VMULT(2,2,NQ,NQ,T4,T4,T41)
000070      CALL TMULT(PV1,TH1,AK,NQ)
000077      DO 300 I=1,NQ
000080      TM(I)=Z(I)-H(I)
000081      IF (ABS(TM(I)).LT.PI) GO TO 350
000083      IF (Z(I).GT.H(I)) T4(I)=TM(I)-TWOPI
000085      30- IF (Z(I).LT.H(I)) TM(I)=TM(I)+TWOPI
000087      35- CONTINUE
000089      CALL VMULT(2,2,NQ,NQ,AK,T4,T41)
000091      XH(1)=XH1(1)+T4(1)
000092      XH(2)=XH1(2)+T4(2)
000094      IF (NI.LE.1) GO TO 110
000096      CALL OHX(XH,DH1,N)
000098      YQ2=2*NQ
000100      DO 400 I=1,NQ2
000102      TM1(I)=OH(I)-OH1(I)
000104      TM2(I)=T4(I)
000106      IF (NO.EQ.1) GO TO 500
000108      TM2(2)=TM1(3)
000110      00- CALL VMULT(2,2,NQ,2,T41,TH2,T4)
000112      A1=0.
000114      DO 600 I=1,NQ2,2
000116      60- A1=A1+TM1(I)
000118      TM1(I)=OH(I)
000120      TM1(2)=OH(2)
000122      IF (NO.EQ.1) GO TO 700
000124      TM1(2)=OH(3)
000126      TM1(4)=OH(4)
000128      70- CALL VMULT(2,2,NQ,2,T41,OH,T4)
000130      A2=0.
000132      DO 800 I=1,NQ2,2
000134      80- A2=A2+T4(I)
000136
000138      KFNLF 367
000139      KFNLF 368
000140      KFNLF 369
000141      KFNLF 370
000142      KFNLF 371
000143      KFNLF 372
000144      KFNLF 373
000145      26MAR71 25
000146      KFNLF 375
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000164      KFNLF 393
000165      26MAR71 26
000166      26MAR71 27
000167      26MAR71 28
000168      26MAR71 29
000169      26MAR71 30
000170      26MAR71 31
000171      KFNLF 395
000172      KFNLF 396
000173      KFNLF 397
000174      26MAR71 31
000175      KFNLF 398
000176      KFNLF 399
000177      KFNLF 400
000178      KFNLF 401
000179      KFNLF 402
000180      KFNLF 403
000181      KFNLF 404
000182      KFNLF 405
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000184      KFNLF 407
000185      KFNLF 408
000186      KFNLF 409
000187      KFNLF 410
000188      KFNLF 411
000189      KFNLF 412
000190      KFNLF 413
000191      KFNLF 414
000192      KFNLF 415
000193      KFNLF 416
000194      KFNLF 417
000195      KFNLF 418
000196      KFNLF 419

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000274 IF (K,GE,N1) GO TO 1100
000302 IF (A1,LE,EP5,A2) GO TO 1100
000306 DO 970 I=1,N02
000319 900 DMIT=DMIT(I)
000320 1000 CONTINUE
000323 1100 CALL OFX(XH,OF,N)
000326 CALL FX(XH,F,N)
000334 CALL VMULT(2,2,NQ,2,AK,OH,TH)
000343 GO TO 1350
000347 1300 CONTINUE
000347 TH1(1)=1.-TH(1)
000351 TH1(2)=TH(3)
000352 TH1(3)=TH(2)
000354 TH1(4)=1.-TH(4)
000356 CALL FTAF(TH1,PN1,PH,2,2)
000361 TH1(7)=AK(1)
000362 TH1(2)=AK(2)
000364 IF (NQ,20,1) GO TO 1200
000372 TH1(2)=AK(3)
000373 TH1(3)=AK(2)
000374 TH1(4)=AK(4)
000376 1200 CALL FTAF(TH1,2,TH,NQ,2)
000402 DO 1300 I=1,3
000412 1301 PN(I)=PN(I)+TH(I)
000424 1350 PN(1)=(1.-TH(1))*PN1(1)+TH(3)*PN1(2)
000427 PN(2)=(1.-TH(1))*PN1(2)+TH(3)*PN1(3)
000432 PN(3)=TH(2)*PN1(2)+(1.-TH(4))*PN1(3)
000440 CALL FX(XH,XL1,N)
000445 XH1(1)=F(1)
000446 XH1(2)=F(2)
000450 TH1(1)=OF(1)
000451 TH1(2)=OF(3)
000452 TH1(3)=OF(2)
000454 TH1(4)=OF(4)
000456 CALL FTAF(TH,PN1,PL1,2,2)
000461 CALL FTAF(TH,PN,PN1,2,2)
000471 DO 1400 I=1,3
000501 PL1(I)=PL1(I)+A(I)
000502 1400 PN1(I)=PN1(I)+A(I)
000503 RETURN
000505 END

```

SUBPROGRAM LENGTH
000600

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

100	-	000140	200	-	000057	300	-	000124	350	-	000137
500	-	000212	700	-	000252	1100	-	000324	1200	-	000377
1301	-	000350	1350	-	000421						

BLOCK NAMES AND LENGTHS

CM1	-	000112	CM3	-	000786	CM4	-	000304	CM6	-	000002
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VARIABLE ASSIGNMENTS

A	-	000.03C02	AK	-	000524	A1	-	000575	A2	-	000576
OF	-	000530	DH	-	000534	DH1	-	000540	GPS	-	000577
F	-	000544	H	-	000546	I	-	000573	K	-	000571
NT	-	000572	N7	-	000511C01	N02	-	000574	NT	-	000571
PI	-	000001C04	PV	-	000550	R	-	000000C02	TM	-	000553
TM1	-	000557	T42	-	000563	TWOPI	-	000001C04	X4	-	000557
XT	-	000.03C03	Z	-	000020C3						

START OF CONSTANTS

000510

START OF TEMPORARIES

000513

START OF INDIRECTS

000921

UNUSED COMPILER SPACE

035000

```

000007 SUBROUTINE FTA(F,A,C,NR,NC)
      DIMENSION F(1), A(1), C(1)
      COMPUTES C=F(TRANPOSE)*A WHERE A IS SYMMETRIC AND STORED
      IN TRIANGULAR FORM
      C(1)=F(1)*A(1)
      IF (NR.GT.1) GO TO 1)
      IF (NR.LE.1) RETURN
      C(1)=C(1)+F(2)*A(2)
      C(2)=F(1)*A(2)+F(2)*A(3)
      RETURN
000026 1) IF (NR.GT.1) GO TO 2)
      C(2)=F(2)*A(1)
      RETURN
000031 2) C(1)=C(1)+F(2)*A(2)
      C(2)=F(3)*A(1)+F(4)*A(2)
      C(3)=F(1)*A(2)+F(2)*A(3)
      C(4)=F(3)*A(2)+F(4)*A(3)
      RETURN
000057 END
000060

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SUBPROGRAM LENGTH
 000101

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS
 10 - 000027 20 - 000034

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS

START OF CONSTANTS
 000063

START OF TEMPORARIES
 000064

START OF INDIRECTS
 000066

UNUSED COMPILER SPACE
 037000

```

000007      SUBROUTINE FTAF(F,A,C,NR,NC)
000008      DIMENSION F(1), A(1), C(1)
000009      EVALUATES F(TRANPOSE) A F
000010      F HAS NR ROWS AND NC COLUMNS
000011      C(1)=F(1)*A(1)
000012      IF (NR.GT.1) GO TO 10
000013      IF (NC.LE.1) RETURN
000014      C(2)=F(1)*A(1)
000015      C(3)=F(2)*A(1)
000016      RETURN
000017      1. C(1)=C(1)+(2.*A(2)*F(1)+A(3)*F(2))*F(2)
000018      IF (NC.LE.1) RETURN
000019      C1=A(1)*F(3)+A(2)*F(4)
000020      C2=A(2)*F(3)+A(3)*F(4)
000021      C(2)=C1*F(1)+C2*F(2)
000022      C(3)=C1*F(3)+C2*F(4)
000023      RETURN
000024      END
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SUSPROGRAM LENGTH
000103

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS
10 - 000'25

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS
C1 - 00C191 C2 - J30102

START OF CONSTANTS
000062

START OF TEMPORARIES
000064

START OF INDIRECTS
000079

UNUSED COMPILER SPACE
G37000

```

000005 SUBROUTINE FX(X,C,N)
000005 DIMENSION X(1), C(1), T(2)
000005 COMMON / CN1/ FL, SL, HL, XLF(4)
000005 LOGICAL FL, SL, HL
000005 IF (FL) GO TO 13
000005 C NONLINEAR DYNAMICS COMPUTED HERE
000005 C IDENTITY LINEAR IS SPECIAL CASE
000006 C(1)=X(1)
000007 C(2)=X(2)
000011 RETURN
000012 1) T(1)=XLF(1)*X(1)+XLF(3)*X(2)
000015 T(2)=XLF(2)*X(1)+XLF(4)*X(2)
000020 C(1)=T(1)
000021 C(2)=T(2)
000022 RETURN
000023 ENTRY DFX
000033 IF (FL) GO TO 23
000033 C NONLINEAR GRADIENT COMPUTED HERE
000033 C IDENTITY IS SPECIAL CASE
000035 C(1)=1.
000036 C(2)=0.
000037 C(3)=0.
000040 C(4)=1.
000041 RETURN
000042 2) C(1)=XLF(1)
000044 C(2)=XLF(2)
000045 C(3)=XLF(3)
000047 C(4)=XLF(4)
000050 RETURN
000051 END

```

SUBPROGRAM LENGTH
000067

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS
10 - 00012 - 20 - 00043

BLOCK NAMES AND LENGTHS
C00 - 000207

VARIABLE ASSIGNMENTS
DFX - 00064 - FL - 000000C01 HL - 000020C01 SL - 000001C01
T - 00065 - XLF - 000003C01

START OF CONSTANTS
000054

START OF TEMPORARIES
000056

START OF INDIRECTS
000060

UNUSED COMPILER SPACE

492 KFNL F
493 KFNL F
494 KFNL F
495 KFNL F
496 KFNL F
497 KFNL F
498 KFNL F
499 KFNL F
500 KFNL F
501 KFNL F
502 KFNL F
503 KFNL F
504 KFNL F
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KFNLF 522
 KFNLF 523
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 KFNLF 538
 KFNLF 539
 KFNLF 540
 KFNLF 541
 KFNLF 542
 KFNLF 543

FUNCTION GRN01(JM)
 COMMON / CM6/ PI, TWOPI
 COMMON / CM13/ ISI, IR(2), X(2)
 DATA (IM=349755813687), (JM=2513271)
 J=IR(1)
 IF (NM.NE.J) GO TO 10
 GO TO (22,40), J
 1. IR(2)=ISI
 2. IR(1)=2
 3. IR(1)=1,2
 IR(2)=400(IR(2)*JM,IM)
 3. X(1)=FLOAT(IR(2))/FLOAT(IM)
 X1=SQRT(CABS(-2.*ALOG(X(1))))
 X(2)=TWOPI*X(2)
 X(1)=X1*SIN(X(2))
 X(2)=X1*COS(X(2))
 GRN01=X(1)
 RETURN
 4. IR(1)=1
 GRN01=X(2)
 RETURN
 5. GRN01=0
 END

SUBPROGRAM LENGTH
 000126

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS
 10 - 000015 20 - 000015 40 - 000057

BLOCK NAMES AND LENGTHS
 CM6 - 000002 CM10 - 000005

VARIABLE ASSIGNMENTS
 GRN01 - 000124 I - 000124 IM - 000121 IR - 000001002
 ISI - 000002 J - 000123 JM - 000122 TWOPI - 000001001
 X - 000002 X1 - 000125

START OF CONSTANTS
 000064

START OF VARIABLES
 000066

START OF INDIRCTS
 000111

UNUSED COMPILER SPACE
 037000


```

VARIABLE ASSIGNMENTS
A - 00C164 B - 0016J REN - J00161 GEN - 00C162
O - 000163 OHX - 000155 FL - 00C003CU3 HL - 00C02C07
NQ - 00011C01 MT - 00011C01 SEN - 000163 SL - 00C01C07
Y - 00C155 TAOPI - 00001C02 XL - 00C03CU3 ALH - 00C13C07

START OF CONSTANTS
000131

START OF TEMPORARIES
000133

START OF INDIRECTS
000150

UNUSED COMPILER SPACE
036500

```


SUBPROGRAM LENGTH
000277

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS					
20	- 000153	30	- 000163		
BLOCK NAMES AND LENGTHS					
CM1	- 000305	CM2	- 000333		
CM6	- 000311	CM7	- 000083		
VARIABLE ASSIGNMENTS					
A	- 00003003	AI	- 000261	AI1	- 00003001
A22	- 00003001	I	- 000264	IX2	- 00003001
MLF	- 00017035	WS	- 000266	NN	- 00003001
NY	- 00001001	PC	- 000272	PN	- 00002502
P11	- 000271	P12	- 000272	P22	- 000273
RLAHN2	- 00022032	RT	- 000274	RLAHN1	- 00014002
S	- 000253	SP11	- 000274	RUN2	- 00004002
X	- 00001004	AC	- 000267	SP12	- 000275
YC	- 000271	YH	- 00023032	X4	- 00001006
				XT	- 000253
				YN2	- 00000002

START OF CONSTANTS

000220

START OF TEMPORARIES

000231

START OF INDIRECTS

000246

UNUSED COMPILER SPACE

036000

*****VARIABLE GIVEN CONFLICTING TYPES

000112,000112,000112

```

000014 SUBROUTINE INPUT(MC,N,IX1,IX2,EJ,H,JC,H1,H2,H3)
000015 EAL EJ(1), H(1), JC(1), H1(1), H2(1), H3(1)
000016 ***** INPUT
000017 C INTEGER IX1(1), IX2(1)
000018 INPUTS ALL VARIABLES FROM CARDS AND/OR FORMATTED TAPE
000019 ***** INPUT
000020 COMMON / CH1 / N, NSD, NMAX, AT(3), NPTS, NP1, NP, N2, SCANX
000021 COMMON / CH2 / YN2(2), RLANV2(2), PUN2(4), Y1(2), RLANV1(2),
000022 1 RUN1(4), YN1(3), YN2(3), PN(3)
000023 COMMON / CH3 / R(3), A(3), Z(3), SQIR(3), SQIR(3)
000024 COMMON / CH4 / X(2), Z(2), J(2), V(2), XC(2), XK(4)
000025 COMMON / CH5 / PI, TWOPI, NSDC, NMC, NMAX, TS, PS, RST, MLF, CP
000026 COMMON / CH7 / N4, X4(20), PC(30)
000027 COMMON / CH8 / FL, SL, HL, XLF(4), XLS(4), XLH(4)
000028 COMMON / CH9 / E4C(200), V4C(300)
000029 COMMON / CH10 / IST, IZ(2), XZ(2)
000030 LOGICAL SCANX, TS, PS, RST, FL, SL, HL, MLF, CP
000031 INTEGER RE, PR, TP
000032 DATA (RE=5), (PR=6), (TP=7)
000033 READ (RE,51020)
000034 READ (RE,51030)
000035 50001 FORMAT(80H
000036 1
000037 READ (RE,51100) NMC, TS, PS, RST, MLF, CP
000038 50101 FORMAT(I5,3L5)
000039 IF (.NOT.MLF) GO TO 1C
000040 TS=.TRUE.
000041 1 WRITE (PR,50.00)
000042 60001 ORMAT(1H1)//////47X,28HGENERALIZED TWO-DIMENSIONAL /47X,
000043 1 28HDISCRETE NONLINEAR PREDICTOR //47X,
000044 2 24HCAPTAIN KEYNEITH O. SENNE /43X,
000045 3 35HFRANK J. SEILER RESEARCH LABORATORY /48X,
000046 4 25HSAIR FORCE SYSTEMS COMMAND /45X,
000047 5 32HUS. AIR FORCE ACADEMY, CO 80840 /53X,
000048 6 13HFEbruary 1971 /141)
000049 WRITE (PR,50000)
000050 IF (PR) GO TO 1010
000051 C BEGIN CARD INPUT
000052 5 READ (RE,51200) A, NMAX, NM, NSD, NSDC, NP, YQ, IST, FL, SL, HL
000053 50201 FORMAT(I5,I5,3L5)
000054 IF (NMAX.GT.100) NMAX=100
000055 IQ=1
000056 IF (NP.GT.1) IQ=3
000057 READ (RE,50300) (Q(I),I=1,IQ)
000058 50301 FORMAT(8F1J.3)
000059 IR=1
000060 IF (NQ.GT.1) IR=3
000061 READ (RE,50300) (R(I),I=1,IR)
000062 IM=2*N4
000063 IP=IM+M
000064 READ (RE,51300) (X4(I),I=1,IM), (PC(I),I=1,IP)
000065 IR=IP+1
000066 IQ=IQ+1
000067 IF (FL) READ (RE,50300) (XLF(I),I=1,4)
000068 IF (SL) READ (RE,50300) (XLS(I),I=1,IO)
000069 IF (HL) READ (RE,50300) (XLH(I),I=1,IR)
000070 IF (TS) GO TO 101
000071 WRITE (TP) 4, NMAX, NM, NSD, NSDC, NP, HQ, IST, FL, SL, HL,
000072 1 Q, R, XLF, XLS, XLH, NM, PC

```

```

000326 7000: FORMAT(10I20,3L20,58E20.14)
000332 ENO FILE TP
000334 BACKSPACE TP
C BEGIN COMMON FILE
000336 10J PI=4.*ATAN(1.)
000342 THOPI=2.*PI
000343 YMAX=2*M+1
000345 00 200 I=1,MMAX
000360 JC(I)=FLOAT(I-1)*NSD/FLOAT(M)
000366 SQR(I)=SQR(JC(I))
000371 IF (NP.EQ.1) GO TO 300
000377 SQR(2)=Q(2)/SQR(I)
000400 SQR(3)=SQR(Q(3)-SQR(2)*SQR(2))
000410 30J SQR(1)=SQR(I)
000413 IF (NQ.EQ.1) GO TO 400
000421 SQR(2)=Q(2)/SQR(1)
000422 SQR(3)=SQR(Q(3)-SQR(2)*SQR(2))
000432 40J IF (TS) GO TO 50J
000434 WRITE (TP) PI, THOPI, MMAX, SQR(I), SQR(2), SQR(3)
000454 7010: FORMAT(2E20.14,120,57E20.14)
000460 ENO FILE TP
000462 BACKSPACE TP
C ZERO MONTE CARLO TABLE
50J N=J
000464 NC=0
000465 00 600 I=1,200
000466 MC(I)=0.
000472 60J VMC(I)=0.
000473 00 700 I=211,300
000474 VMC(I)=0.
000504 70J VMC(I)=0.
000506 IF (.NOT.TS) CALL MCRITE(MC)
000514 MC=1
C PRINT PROBLEM PARAMETERS
000521 80J IF (TS) WRITE (PR,601'U)
000532 6010: FORMAT(/51X,22HTAPE OUTPUT SUPPRESSED /)
000532 IF (PS) WRITE (PR,602'U)
000543 6020: FORMAT(/45X,31HSAMPLE PATH PRINTOUT SUPPRESSED /)
000543 IF (MLF) WRITE (PR,60250)
000554 6025: FORMAT(/45X,32HLINEARIZED PREDICTOR MONTE CARLO )
000554 WRITE (PR,6030) M4C, YMAX, M, NSD, NSDC, NP, NQ, IST
000600 6030: FORMAT(/52X,18HP208LEY PARAMETERS ///21X,
1 19HTM11UM YMONTE CARLO, 33X,16HSAMPLES PER PATH ///28X,15
2 ///16X,3MGRI0 SIZE, 29X,15HNO. GRI0 SIGMAS, 18X,
3 274'0. TRACKING ELLIPSE SIGMAS///17X,3HM =,15,31X,5HNSD =,15,
4 30X,6HNSDC =,15///13X,16HNUMBER OF INPUTS, 24X
5 17HNUMBER OF OUTPUTS, 21X,21HINITIAL RANDOM NUMBER ///16X,4HNP =
6 ,15,31X,4HNP =,15,29X,115///)
000604 IF (NP.EQ.1) WRITE (PR,60400) Q(1)
000617 6040: FORMAT(54X,13HPROCESS NOISE /54X,13HCOVARIANCE 1 ///51X,2F10.3))
000617 IF (NP.NE.1) WRITE (PR,60400) Q(1), Q(2), Q(3)
000641 IF (NQ.EQ.1) WRITE (PR,60500) R(1)
000655 6050: FORMAT(/52X,17HMEASUREMENT NOISE /54X,13HCOVARIANCE 2 //
1 (51X,2F10.3))
000655 IF (NQ.NE.1) WRITE (PR,60500) R(1), R(2), R(3)
000677 WRITE (PR,50600)
000703 6060: FORMAT(/71X,17HLINEAR PARAMETERS)
000707 IP=2*NP
000710 IQ=2*NQ

```

KFNLF 679
 KFNLF 680
 KFNLF 681
 KFNLF 682
 KFNLF 683
 KFNLF 684
 KFNLF 685
 KFNLF 686
 KFNLF 687
 KFNLF 688
 KFNLF 689
 KFNLF 690
 KFNLF 691
 KFNLF 692
 KFNLF 693
 KFNLF 694
 KFNLF 695
 KFNLF 696
 74AP71 33
 KFNLF 698
 KFNLF 699
 KFNLF 700
 KFNLF 701
 KFNLF 702
 KFNLF 703
 KFNLF 704
 KFNLF 705
 KFNLF 706
 KFNLF 707
 KFNLF 708
 KFNLF 709
 KFNLF 710
 KFNLF 711
 KFNLF 712
 KFNLF 713
 KFNLF 714
 KFNLF 715
 26MA871 50
 26MA871 51
 KFNLF 716
 KFNLF 717
 KFNLF 718
 KFNLF 719
 KFNLF 720
 KFNLF 721
 KFNLF 722
 KFNLF 723
 KFNLF 724
 KFNLF 725
 KFNLF 726
 KFNLF 727
 KFNLF 728
 KFNLF 729
 KFNLF 730
 KFNLF 731
 KFNLF 732
 KFNLF 733
 KFNLF 734

```

000711 IF (FL) WRITE (PR,6070) (XLF(I),I=1,4)
000725 6070: FORMAT(/45X,1HF,9X,2F10.3/55X,2F10.3)
000725 IF (SL) WRITE (PR,6070) (XLS(I),I=1,IP)
000741 6080: FORMAT(/45X,5HSIC4A,7X,2F10.3/55X,2F10.3)
000741 IF (HL) WRITE (PR,6070) (XLH(I),I=1,IO)
000755 6090: FORMAT(/45X,14H,9X,2F10.3/55X,2F10.3)
000755 WRITE (PR,51700)
000761 6100: FORMAT(1H1)
000765 RETURN
C TAPE INPUT BEGINS HERE
000765 100: READ (TP) M, NMAX, NI, NSD, NSDC, NP, N3, IST, FL,
001034 1 SL, HL, O, R, XLF, XLS, XLH, XM, PC
001043 IF (EOF,TP) 50,1313
001043 101: READ (TP)
001063 1 (JC(I),I=1,MHAX)
001072 IF (EOF,TP) 103,1029
001072 102: IS=1
001073 105: IP=NMC+1
001075 IN=NMAX+1
001077 10 140: I=IS,IP
001101 4C=I-1
001102 IF (MCREAD(MC)) 1100,1200,1200
001110 110: IF (MC) 50,500,1700
001112 120: MC=I
001113 10 140: J=1,IN
001115 N=J-1
001116 IF (MCREAD(MC,N,M,EJ,IX,IX2)) 1500,1300,1300
001126 130: IF (N.EQ.0) CALL MCAV(MC,N,X,YN,EMC,VHC)
001136 140: CONTINUE
001143 IS=IP+1
001144 VHC=NMC+16
001146 50 TO 1050
001147 150: IF (N.EQ.0) 60 TO 1713
001150 CALL MCAV(MC,N,X,YN,EMC,VHC)
001153 VI=N+1
001160 60 TO 1800
001161 170: HI=N
001162 180: WRITE (PR,51100) MC, N:
001172 6110: FORMAT(/32X,31HTAPE PESTART ON MONTE CARLO NO. ,I5,
001176 1 16H WITH SAMPLE NO. ,I5/)
001176 60 TO 300
001176 END

```

SUBPROGRAM LENGTH

001532

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

10	-	000751	57	-	001371	103	-	000737	300	-	000411
400	-	000433	500	-	100465	800	-	000522	170	-	000766
1010	-	001344	1727	-	731373	105	-	001074	1103	-	001111
1200	-	001113	1300	-	011127	140	-	001137	1503	-	001150
1700	-	001162	1300	-	001163	500.0	-	001202	53130	-	001214
50230	-	001271	50330	-	001274	600.0	-	001216	60130	-	001315
60200	-	001322	80250	-	001330	60330	-	001336	63470	-	001412

60500 - 001422 61600 - 001423 60700 - 001440 60800 - 001445
 60900 - 001452 61000 - 001457 61100 - 001461 70000 - 001302
 70100 - 001311

BLOCK NAMES AND LENGTHS

CM1 - 000113 CM2 - 000031 CM3 - 000017 CM4 - 000016
 CM6 - 000112 CM7 - 000063 CM8 - 000017 CM9 - 000016
 CM10 - 000005

VARIABLE ASSIGNMENTS

A - 00003003 AT - 00003001 CP - 000011005 EMC - 000000010
 FL - 00000007 HL - 00000207 H1 - 000001 H2 - 000002
 H3 - 000003 I - 001522 IH - 001524 IN - 001527
 IP - 001525 IQ - 001521 IR - 001523 IS - 001526
 IST - 00000011 IZ - 00000101 J - 001530 JC - 000000
 JH - 00000001 MLF - 00001005 HMAX - 00000201 NM - 000000005
 NMAX - 00000405 N4C - 00000305 NP - 00001101 NQ - 00001101
 NSD - 00000101 NSDC - 00000205 N1 - 00002502 P1 - 00002500
 PI - 00000005 PH - 00000603 P - 00000700 RE - 001516
 PS - 00000005 Q - 00000202 RST - 00000700 RUN1 - 00001400
 RLANH1 - 00001202 RLANH2 - 00001201 SL - 00000107 S1TQ - 00001100
 RUN2 - 00000402 SCANX - 001523 TS - 00000505 THOPI - 00000100
 SOTR - 00001403 TP - 00000604 VMC - 000010010 X - 0000000004
 U - 00000404 V - 00001204 XLF - 00000307 XH - 00001307
 XC - 00001004 XK - 00000106 XZ - 00000701 YN - 00002300
 XLS - 00007007 XM - 00000106 XZ - 00000701 YN - 00002300
 YN1 - 00001102 YN2 - 00000702 Z - 00000204

START OF CONSTANTS

001201

START OF TEMPORARIES

001473

START OF INDIRECTS

001510

UNUSED COMPILER SPACE

032300

*****VARIABLE GIVEN CONFLICTING TYPES

000115,000015

```

00010 SUBROUTINE MCAV(MC,N,X,YN,EYC,VHC)
00011 DIMENSION X(1), YN(1), EMC(1), VHC(1), Y(2)
00012 N1= 2*(N-1)
00013 N2= 3*(N-1)
00014 Y(1)=X(1)-YN(1)
00015 Y(2)=X(2)-YN(2)
00016 EMC(N1+1)=CUMAVG(EYC(N1+1),Y(1),MC)
00017 EMC(N1+2)=CUMAVG(EMC(N1+2),Y(2),MC)
00018 VHC(N2+1)=CUMAVG(VHC(N2+1),Y(1)*Y(1),MC)
00019 VHC(N2+2)=CUMAVG(VHC(N2+2),Y(1)*Y(2),MC)
00020 VHC(N2+3)=CUMAVG(VHC(N2+3),Y(2)*Y(2),MC)
00021 RETURN
00022 END

```

```

KFNLF 777
KFNLF 778
KFNLF 779
KFNLF 780
KFNLF 781
KFNLF 782
KFNLF 783
KFNLF 784
KFNLF 785
KFNLF 786
KFNLF 787
KFNLF 788
KFNLF 789

```

```

SUBPROGRAM LENGTH
00015

```

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS

```

N1 - 00013 N2 - 000154 Y - 000151

```

START OF CONSTANTS

```

00014

```

START OF TEMPORARIES

```

00015

```

START OF INDIRECTS

```

00016

```

```

UNUSED COMPILER SPACE
036700

```

```

000002      FUNCTION 'MCREAD(MC)
000002      COMMON / CM9/ E'IC(230), VMC(320)
000002      INTEGER TP
000002      DATA (TP=77)
000002      NRH=0
000002      READ (TP)
000003      READ (TP)
000016      7000) FORMAT(3I2J,20E20.14)
000017      IF (EOF,TP) 40,10
000022      1. NRH=NRH+1
000024      IF (N.NE.-1.OR.MC.NE.MCH.OR.NR.NE.NRH) GO TO 50
000036      READ (TP)
000051      IF (EOF,TP) 43,23
000055      2. NRH=NRH+1
000057      IF (N.NE.-1.OR.MC.NE.MCH.OR.NR.NE.NRH) GO TO 50
000071      READ (TP)
000104      IF (EOF,TP) 40,30
000110      3. NRH=NRH+1
000112      IF (N.NE.-1.OR.MC.NE.MCH.OR.NR.NE.NRH) GO TO 50
000124      MCREAD=1
000125      RETURN
000125      4. BACKSPACE TP
000127      5. MCREAD=-1
000130      6. IF (NRH.EQ.0) RETURN
000133      BACKSPACE TP
000135      NRH=NRH-1
000137      GO TO 60
000137      END

```

SUBPROGRAM LENGTH

000164

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

10	-	000023	23	-	000056	30	-	000111	40	-	000126
50	-	000133	6	-	000131	70000	-	000144			

BLOCK NAMES AND LENGTHS

CM9 - 000764

VARIABLE ASSIGNMENTS

CMC	-	00000001	I	-	000163	MCH	-	000150	MCREAD	-	000155
N	-	000131	NR	-	000162	NRH	-	000157	TP	-	000156
VMC	-	00031301									

START OF CONSTANTS

000142

START OF TEMPORARIES

000147

START OF INDIRECTS

000155

UNUSED COMPILER SPACE

036600

KFNLF 790
 KFNLF 791
 KFNLF 792
 KFNLF 793
 KFNLF 794
 7MAR71 35
 KFNLF 796
 KFNLF 797
 KFNLF 798
 KFNLF 799
 7MAR71 36
 KFNLF 801
 KFNLF 802
 KFNLF 803
 7MAR71 37
 KFNLF 805
 KFNLF 806
 KFNLF 807
 KFNLF 808
 KFNLF 809
 KFNLF 810
 KFNLF 811
 KFNLF 812
 KFNLF 813
 KFNLF 814
 KFNLF 815
 KFNLF 816


```

000002 SUBROUTINE MCRITE(MC)
000002 COMMON / CH9/ EMC(270), VMC(300)
000002 INTEGER TP
000002 DATA (TP=7)
000002 NR=1
000003 WRITE (TP) MC, N, NR, EMC
000004 7000J FORMAT(3I2J,247E20.14)
000005 END FILE TP
000006 BACKSPACE TP
000007 NR=2
000008 WRITE (TP) MC, N, NR, (VMC(I), I=1,150)
000009 END FILE TP
000010 BACKSPACE TP
000011 NR=3
000012 WRITE (TP) MC, N, NR, (VMC(I), I=151,300)
000013 END FILE TP
000014 BACKSPACE TP
000015 RETURN
000016 END

```

```

KFNL F 817
KFNL F 818
KFNL F 819
KFNL F 820
KFNL F 821
KFNL F 822
74AR71 38
KFNL F 824
KFNL F 825
KFNL F 826
KFNL F 827
74AR71 39
KFNL F 829
KFNL F 830
KFNL F 831
74AR71 40
KFNL F 833
KFNL F 834
KFNL F 835
KFNL F 836

```

```

SUBPROGRAM LENGTH
000107

```

```

FUNCTION ASSIGNMENTS

```

```

STATEMENT ASSIGNMENTS
70000 - 000374

```

```

BLOCK NAMES AND LENGTHS
CH9 - 000764

```

```

VARIABLE ASSIGNMENTS

```

```

EMC - 000:00091 I - 000106 N - 000104 NR - 000105
TP - 000103 VMC - 000103.1

```

```

START OF CONSTANTS
000072

```

```

START OF TEMPORARIES
000077

```

```

START OF INDIRECTS
000103

```

```

UNUSED COMPILER SPACE
037000

```

KFNLF 837
 KFNLF 838
 KFNLF 839
 KFNLF 840
 KFNLF 841
 KFNLF 842
 KFNLF 843
 KFNLF 844
 KFNLF 845
 KFNLF 846
 KFNLF 847
 KFNLF 848
 KFNLF 849
 KFNLF 850
 KFNLF 851
 KFNLF 852

C SUBROUTINE MEAS(N)
 GENERATES MEASURED OUTPUT VARIABLES
 COMMON / CH1/ IT(3), AT(3), IT1(3), NQ
 COMMON / CH3/ IM(12), SQTR(3)
 COMMON / CH4/ X(2), Z(2), U(2), V(2)
 DIMENSION H(2)
 DO 10 I=1,NQ
 1) V(I)=GRN01(0)
 VT=V(I)
 V(I)=SQTR(1)*V(I)
 IF (NQ.GT.1) V(2)=SQTR(2)*VT+SQTR(3)*V(2)
 CALL HX(X,H,N)
 DO 20 I=1,NQ
 2) Z(I)=H(I)+V(I)
 RETURN
 END
 000040

SUBPROGRAM LENGTH
 000156

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLOCK NAMES AND LENGTHS
 CH1 - 000012 CH3 - 000017 CH4 - 000010

VARIABLE ASSIGNMENTS

AT - 000033C01 H - 000054 IT - 000000C01
 IT1 - 000006C01 NQ - 000014C02 IM - 000000C02
 U - 000004C03 V - 000000C02
 Z - 000002C03 VT - 000000C02

START OF CONSTANTS
 000043

START OF TEMPORARIES
 000045

START OF INDIRECTS
 000051

UNUSED COMPILER SPACE
 037000

853 K=VLF
 854 KFNLF
 855 KFNLF
 856 KFNLF
 857 KFNLF
 858 KFNLF
 859 KFNLF
 860 KFNLF
 861 KFNLF
 862 KFNLF
 863 KFNLF
 864 KFNLF
 865 KFNLF
 866 KFNLF
 867 KFNLF
 868 KFNLF
 869 KFNLF
 870 KFNLF
 871 KFNLF
 872 KFNLF
 873 KFNLF
 874 KFNLF
 875 KFNLF
 876 KFNLF
 877 KFNLF
 878 KFNLF

```

000005 C SUBROUTINE MINV(A,AI,N)
          DIMENSION A(1), AI(1)
          COMPUTES ACTUAL OR MOORE-PENROSE PSEUDO INVERSE OF A
          IF (N.GT.1) GO TO 10
          AI(1)=0.
          IF (ABS(A(1)).GT.1.E-40) AI(1)=1./A(1)
          RETURN
          10 O=A(1)*A(3)-A(2)*A(2)
          IF (ABS(O).LT.1.E-40) GO TO 20
          AT=A(3)/O
          AI(1)=AT
          RETURN
          20 O=ABS(A(1)*A(3))
          IF (O.GT.1.E-40) GO TO 30
          AI(1)=0.
          AI(2)=O.
          AI(3)=O.
          RETURN
          30 O=O*O
          AI(1)=A(1)/O
          AI(2)=A(2)/O
          AI(3)=A(3)/O
          RETURN
          END
000052
000052

```

SUBPROGRAM LENGTH
 000077

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS
 10 - 000317 20 - 000334 30 - 000045

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS
 AT - 000376 0 - 000075

START OF CONSTANTS
 000055

START OF TEMPORARIES
 000060

START OF INDIRECTS
 000064

UNUSED COMPILER SPACE
 037000

```

000005 REAL FUNCTION NORMAL(X,XM,C)
000005 DIMENSION X(1), XM(1), C(1)
000006 Y1=X(1)-XM(1)
000006 Z=X(2)-XM(2)
000010 NORMAL=EXP(-(Y1*(1*C(1)+2.*(1*(2*C(2)+Y2*(2*C(3))/2.))
000025 RETURN
000025 END

```

KFNLF 879
 KFNLF 880
 KFNLF 881
 KFNLF 882
 KFNLF 883
 KFNLF 884
 KFNLF 885

SUBPROGRAM LENGTH
 000054

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS

NORMAL 000051 Y1 - 000052 Y2 - 000053

START OF CONSTANTS
 000030

START OF TEMPORARIES
 000033

START OF INDIRECTS
 000042

UNUSED COMPILER SPACE
 037100

944 KFNL F
 945 KFNL F
 946 KFNL F
 947 KFNL F
 948 KFNL F
 949 KFNL F
 950 KFNL F
 951 KFNL F
 952 KFNL F
 953 KFNL F
 954 KFNL F
 955 KFNL F
 956 KFNL F
 957 KFNL F
 958 KFNL F
 959 KFNL F
 960 KFNL F
 961 KFNL F
 962 KFNL F
 45 7MAR71
 964 KFNL F
 965 KFNL F
 966 KFNL F

000404 110 NRH=NRH+1
 000406 IF (N.NE.NH.OR.MC.NE.MCH.OR.NR.NE.NRH) GO TO 140
 000420 12J IF (NBS.NE.0) NREAD=1
 000422 CALL SXTX(S,N)
 000425 CALL AX(S,A,N)
 000434 CALL MINV(A,AI,2)
 000437 A11=-.5*AI(1)
 000441 A12=-AI(2)
 000442 A22=-.5*AI(3)
 000444 RETURN
 000446 130 BACKSPACE TP
 000450 14J NREAD=-1
 000451 15J IF (NRH.LE.0) GO TO 16J
 000453 BACKSPACE TP
 000455 NRH=NRH-1
 000457 GO TO 150
 000457 160 IF (N.EQ.0) RETURN
 000462 IF (NBS.EQ.0) GO TO 10
 000463 BACKSPACE TP
 000465 READ (TP) MC, N, NRH
 000476 NBS=0
 000477 GO TO 150
 000503 END

SUBPROGRAM LENGTH

000555

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS
 0 - 00022 2J
 60 - 00026 7
 110 - 00040 12J
 150 - 00045 160

BLOCK NAMES AND LENGTHS
 CH1 - 00013 CM2
 CM5 - 00001 CH10

VARIABLE ASSIGNMENTS

A - 00003003 AI
 A22 - 00005001 C
 J - 000554 MCH
 NBS - 000544 NH
 NPQ - 00007001 NPTS
 NRH - 000345 NT
 NWR - 000542 PN
 RLAH2 - 00002002 RT
 S - 000534 SCANX
 V - 00006004 X
 XKN - 00033002 X7
 YN2 - 00006002 Z

START OF CONSTANTS

000506

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```

000010 SUBROUTINE WRITE(MC,N,W,EJ,IX1,IX2)
000010 DIMENSION W(1), EJ(1), IX1(1), IX2(1)
000010 COMMON / CM1/ NT(2), HMAX, AT(3), NPTS, NPQ, NT1(2), SCANX
000010 COM+JN 7 CM2/ YN2(2), ZLANY2(2), RUM2(2), YN1(2), RLANY1(2),
000010 1 RUN1(4), PN1(3), YN(2), PV(3), XKN(3)
000010 COMMON / CM4/ X(2), Z(2), U(2), V(2), X(2), XK(4)
000010 COMMON / CM5/ C
000010 INTEGER TP
000010 LOGICAL SCANX
000010 DATA (TP=7)
000010 MH=HMAX+HMAX
000010 NWR=MM/247
000015 NLR=MOD(MM,247)
000020 NR=1
000021 WRITE(TP) MC, N, NR, NPTS, NPQ, IZ, XZ, SCANX, C,
000021 1 U, X, V, Z, XC, XCN, YN, YN1, YN2, RLANY1, RLANY2,
000021 2 RUN1, RUM2, PN1, (IX1(I), I=1, HMAX), (IX2(I), I=1, HMAX)
000021 7000J FORMAT(I2,2E20.14,L20,4E20.14,198I20)
000021 END FILE TP
000021 BACKSPACE TP
000021 NR=NR+1
000021 NU=0
000021 IF (NWR.EQ.0) GO TO 2J
000021 DO 1C I=1,NWR
000021 NL=(I-1)*247+1
000021 NU=I*247
000021 WRITE(TP) MC, N, NR, (W(J), J=NL,NU)
000021 7010J FORMAT(3I2,247E20.14)
000021 -NO FILE TP
000021 BACKSPACE TP
000021 1J NR=NR+1
000021 2J IF (NLR.EQ.0) GO TO 3J
000021 NL=NU+1
000021 NU=NU+NLR
000021 WRITE(TP) MC, N, NR, (W(J), J=NL,NU)
000021 END FILE TP
000021 BACKSPACE TP
000021 NR=NR+1
000021 NU=0
000021 IF (NWR.EQ.0) GO TO 5J
000021 DO 4J I=1,NWR
000021 NL=(I-1)*247+1
000021 NU=247*I
000021 WRITE(TP) MC, N, NR, (EJ(J), J=NL,NU)
000021 END FILE TP
000021 BACKSPACE TP
000021 4J NR=NR+1
000021 5J IF (NLR.EQ.0) RETURN
000021 NL=NU+1
000021 NU=NU+NLR
000021 WRITE(TP) MC, N, NR, (EJ(J), J=NL,NU)
000021 END FILE TP
000021 BACKSPACE TP
000021 RETURN
000021 END

```

KFNLF 967
 KFNLF 968
 KFNLF 969
 KFNLF 970
 KFNLF 971
 KFNLF 972
 KFNLF 973
 KFNLF 974
 KFNLF 975
 KFNLF 976
 KFNLF 977
 KFNLF 978
 KFNLF 979
 KFNLF 980
 KFNLF 981
 2:MAR:1 27
 KFNLF 983
 KFNLF 984
 KFNLF 985
 KFNLF 986
 KFNLF 987
 KFNLF 988
 KFNLF 989
 KFNLF 990
 KFNLF 991
 KFNLF 992
 KFNLF 993
 7:AR71 46
 KFNLF 995
 KFNLF 996
 KFNLF 997
 KFNLF 998
 KFNLF 999
 KFNLF 1000
 KFNLF 1001
 7:AR71 47
 KFNLF 1003
 KFNLF 1004
 KFNLF 1005
 KFNLF 1006
 KFNLF 1007
 KFNLF 1008
 KFNLF 1009
 KFNLF 1010
 7:AR71 48
 KFNLF 1012
 KFNLF 1013
 KFNLF 1014
 KFNLF 1015
 KFNLF 1016
 KFNLF 1017
 7:AR71 49
 KFNLF 1019
 KFNLF 1020
 KFNLF 1021
 KFNLF 1022

SUBPROGRAM LENGTH
000377

- FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS
A0 - 000201 30
70100 - 000355

BLOCK NAMES AND LENGTHS
CH1 - 000313 CM2
CH10 - 000305

VARIABLE ASSIGNMENTS

AT	-	000303C01	C	-	000373	IZ	-	000001C05
J	-	000375	HM	-	000102C01	NL	-	000375
NLR	-	000371	NPQ	-	000106C01	NQ	-	000372
NT	-	000303C01	NT1	-	000374	NHR	-	000370
PN	-	000250C02	PN1	-	00012C02	RLAHN1	-	00002C02
RUN1	-	00014C02	RUN2	-	00012C01	IP	-	000366
U	-	00004C03	V	-	00012C03	XC	-	00010C07
XK	-	00012C03	XKN	-	00010C05	YN	-	000123C02
(N1	-	00011C02	(N2	-	00012C03		-	

START OF CONSTANTS

000344

START OF TEMPORARIES

000360

START OF INDIRECTS

000366

UNUSED COMPILER SPACE

035600


```

000002 SUBROUTINE OUTPUT(N)
000002 COMMON / CM1/ NT(6), NPTS
000002 COMMON / CM2/ Y(4), YN1(2), RLAMN1(2), RUN1(4), PV1(3), YN(2),
000002 PH1(3), XKN(3)
000002 COMMON / CM4/ X(2), Z(2), U(2), V(2)
000002 COMMON / CM6/ PI
000002 INTEGER PR
000002 DATA (PR=5)
000002 PHI=ATAN2(RUN1(2),RUN1(1))*180./PI
000002 IF (Y(4).GT.10) GO TO 10
000002 ARITE (PR,50000) N, X(1), YN(1), PV1(1), YN1(1), PN1(1),
000013 1 PH1(2), PLAMN1(1), X(2), YN(2), PH(2), PH(3), YN1(2), PN1(2),
000062 2 PH1(3), RLAMN1(2), PHI
000063 6000 FORMAT(I4,3I,5E10.3,F5.0/)
000064 10 WRITE (PR,50000) N, X(1), YN(1), Z(1), YN(1), PH(1), PN(1),
000064 1 YN1(1), PH1(1), PV1(2), RLAMN1(1), NPTS, V(2), Z(2), U(2),
000064 2 X(2), YN(2), PH(2), PH(3), YN1(2), PN1(2), RLAMN1(2),
000064 3 PH1
000152 60100 FORMAT(I4,1X,11E10.3,I5/5X,11E10.3,F5.0/)
000153 RETURN
000154 END

```

```

KFNL F 1023
KFNL F 1024
KFNL F 1025
KFNL F 1026
KFNL F 1027
KFNL F 1028
KFNL F 1029
KFNL F 1030
KFNL F 1031
KFNL F 1032
KFNL F 1033
KFNL F 1034
KFNL F 1035
KFNL F 1036
KFNL F 1037
KFNL F 1038
KFNL F 1039
KFNL F 1040
KFNL F 1041
KFNL F 1042
KFNL F 1043
KFNL F 1044

```

SUBPROGRAM LENGTH

000203

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

10 - 000159 01000 - 010161 60100 - 000166

BLOCK NAMES AND LENGTHS

CH1 - 000107 C12 - 010133 C14 - 000010 C16 - 000031

VARIABLE ASSIGNMENTS

```

NPTS - 000106C01 NT - 010102C01 PHI - 010102
PH - 000125C02 PH1 - 010121C02 PR - 010101
RUN1 - 000140C02 U - 010104C03 V - 010106C03 X - 010106C03
XKN - 000133C02 Y1 - 010123C02 YN1 - 010102C02 Y2 - 010102C02
Z - 000102C03

```

START OF CONSTANTS

000157

START OF TEMPORARIES

000173

START OF INDIRECTS

000177

UNUSED COMPILER SPACE

036500

```

000005      FUNCTION QUAD(I,J,P)
000005      DIMENSION P(1)
000017      QUAD=(I*P(1)+J*P(2))*I+(I*P(2)+J*P(3))*J
000017      RETURN
000017      END

```

```

KFNL F      1J45
KFNL F      1J46
KFNL F      1J47
KFNL F      1J48
KFNL F      1J49

```

```

SUBPROGRAM LENGTH
000035

```

```

FUNCTION ASSIGNMENTS

```

```

STATEMENT ASSIGNMENTS

```

```

BLOCK NAMES AND LENGTHS

```

```

VARIABLE ASSIGNMENTS
QUAD - 000-34

```

```

START OF CONSTANTS
000022

```

```

START OF TEMPORARIES
000023

```

```

START OF INDIPECTS
000031

```

```

UNUSED COMPILER SPACE
037100

```

```

000007 SUBROUTINE QMKPLI(X,Z,NP,NL)
000007 DIMENSION X(1), Y(1), Z(1), IA(36), IO(119), ZH(991)
000007 INTEGER P2
000007 DATA (P2=6), P2(3=1H)
000007 DATA ((IA(I), I=1, 36)=1HC, 1H1, 1H2, 1H3, 1H4, 1H5, 1H6, 1H7, 1H8, 1H9, 1HA,
1 1H9, 1HC, 1HD, 1HE, 1HF, 1HG, 1HH, 1HI, 1HJ, 1HK, 1HL, 1HM, 1HN, 1HO, 1HP, 1HQ
2 1H3, 1HS, 1HT, 1HU, 1HV, 1HW, 1HX, 1HY, 1HZ)
000007 IF (NL.GT.36) NL=36
000007 ZMIN=Z(1)
000007 ZMAX=Z(1)
000007 K=1
000007 DO 10 J=1, NP
000007 DO 11 I=1, NP
000007 IF (Z(K).GT.ZMAX) ZMAX=Z(K)
000007 IF (Z(K).LT.ZMIN) ZMIN=Z(K)
000007 K=K+1
000007 WRITE (PR, 9000) NL, X(1), X(NP), (1), (NP), ZMIN, ZMAX
000007 6000 FORMAT(1H0, 21HQMKPLI OUTPUT WITH, 15, 7H LEVE-S/1H,
1 20HX LIMITS-HORIZONTAL-, 2E15.7/1H, 18HY LIMITS-VERTICAL-, 2A,
2 2E15.7/1H, 9HZ LIMITS-, 11X, 2E15.7/1H)
000007 IF (ZMAX.LT.1.E-10) RETURN
000007 XNL=NL*1
000007 XNL=ZMAX*EXP(-XNL*XNL/2.)
000007 YH=NP*Y
000007 DO 15 I=1, MM
000007 ZH(I)=0.
000007 15 IF (Z(I).GT.XNL) ZH(I)=SQRT(ABS(2.*ALOG(Z(I)/ZMAX)))-NL-1.
000007 DO 30 J=1, 50
000007 IL=(NP)-(I-1.)*(NP)-(1)/59.
000007 DO 20 J=1, 19
000007 XI=(J-1.)*(X(NP)-X(1))/118.*X(1)
000007 IL=3.*ALAG3(XI, YI, X, Y, ZH, NP)*NL+2.)*+1.
000007 IO(J)=IB
000007 IF (MOD(I, 3).NE.0) GO TO 20
000007 IL=IL/3
000007 IO(J)=IA(IL)
000007 20 CONTINUE
000007 30 WRITE (PR, 60100) IO
000007 60100 FORMAT(1H, 119A1)
000007 WRITE (PP, 60200)
000007 60200 FORMAT(1H1)
000007 RETURN
000007 END

```

SUBPROGRAM LENGTH
002550

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS
10 - 00035 20 - 00231 6000 - 00261 60100 - 000313
60200 - 000316

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS

I	-	002542	IA	-	000342	I3	-	002535	IL	-	002547
IO	-	000405	J	-	002541	K	-	002543	IM	-	002544
PR	-	002534	AI	-	002545	XNL	-	002543	YI	-	002545
ZH	-	000575	ZMAX	-	002537	ZMIN	-	002536			

START OF CONSTANTS

000260

START OF TEMPORARIES

000320

START OF INDIRECTS

000334

UNUSED COMPILER SPACE

035700

```

000006 SUBROUTINE ROTATE(P,PR,U,RL)
000015 DIMENSION P(1), PR(1), U(1), RL(1)
000027 PR(1)=(U(1)*P(1)+U(2)*P(2))*U(1)+(U(1)*P(2)+U(2)*P(3))*U(2)
000041 PR(2)=(U(1)*P(1)+U(2)*P(2))*U(2)+(U(1)*P(3)+U(2)*P(4))*U(3)
000053 PR(1)=RL(1)*PR(1)
000054 PR(2)=SQRT(RL(1)*RL(2))*PR(2)
000064 PR(3)=RL(2)*PR(3)
000067 RETURN
000067 END

```

```

KFNL F 1050
KFNL F 1051
KFNL F 1052
KFNL F 1053
KFNL F 1054
KFNL F 1055
KFNL F 1056
KFNL F 1057
KFNL F 1058
KFNL F 1059

```

```

SUBPROGRAM LENGTH
000134

```

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS

```

START OF CONSTANTS
000072

```

```

START OF TEMPORARIES
000073

```

```

START OF INDIRECTS
000114

```

```

UNUSED COMPILER SPACE
037000

```

```

000004      SUBROUTINE SCALE(M,EJ)
000004      DIMENSION W(1), FJ(1)
000004      C      SCALES THE NEW DENSITY W TO INTEGRATE TO ONE
000004      COMMON / CH1/ M, NSD, MMAX
000004      COMMON / CH2/ AT(10), RLAMN1(2), AT1(7), N(2), PN(3)
000004      COMMON / CH3/ C, Y1, Y2, S11, S12, S22
000006      YN(1)=Y1/C
000007      YN(2)=Y2/C
000011      PN(1)=S11/C-PN(1)*N(1)
000013      PN(2)=S12/C-PN(1)*YN(2)
000015      PN(3)=S22/C-PN(2)*YN(2)
000020      YN=MMAX*MMAX
000022      XS=(C*ISD*NSD*RLAMN1(1)*RLAMN1(2))/(M*Y)
000032      DO 10 I=1, 4M
000040      1. FJ(I)=F(I)/XS
000042      RETURN
000042      END

```

KFNLF 1060
KFNLF 1061
KFNLF 1062
KFNLF 1063
KFNLF 1064
KFNLF 1065
KFNLF 1066
KFNLF 1067
KFNLF 1068
KFNLF 1069
KFNLF 1070
KFNLF 1071
KFNLF 1072
KFNLF 1073
KFNLF 1074
KFNLF 1075
KFNLF 1076

SUBPROGRAM LENGTH
000064

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLOCK NAMES AND LENGTHS

CH1 - 000003 CH2 - 000003 CH3 - 000005

VARIABLE ASSIGNMENTS

AT - 00000002 AT1 - 00001402 C - 00000003 I - 000063
M - 00000003 M1 - 000001 MMAX - 000002001 NSD - 00000001
PN - 00000002 RLAMN1 - 00001202 S11 - 0000000303 S12 - 000004003
S22 - 00000003 XS - 00000002 YN - 000002002 Y1 - 000001003
Y2 - 00000003

START OF CONSTANTS

000045

START OF TEMPORARIES

000047

START OF INDIRECTS

000057

UNUSED COMPILER SPACE

037000

```

000005      SUBROUTINE SHIFT(N,EJ,JC)
000006      EAL JC(1), EJ(1)
000007      SHIFTS AND UPDATES GRID PARAMETERS
000008      ***** SHIFT *****
000009      DIMENSION X(2), F(2), A(3), XP(2), S(4)
000010      COMMON / CM1/ M, NSD, HMAX
000011      COMMON / CM2/ YN2(2), RLANM2(2), YN1(2), RLANM1(2),
000012      1 RUN1(4), PN1(3)
000013      DO 10 I=1,2
000014      10 RLANM2(I)=RLANM1(I)
000015      10 RUN2(3)=RUN1(3)
000016      10 RUN2(4)=RUN1(4)
000017      10 26 I=1,2
000018      10 YN1(I)=0.
000019      20 PN1(I)=C.
000020      20 PN1(3)=0.
000021      K=1
000022      DO 40 J=1,HMAX
000023      40 XP(2)=JC(I)
000024      40 40 I=1,HMAX
000025      40 XP(1)=JC(I)
000026      CALL XUNPRM(X,2,RUN2,RLANM2,YN2,X)
000027      CALL FAX(F,N)
000028      CALL SX(X,S,N)
000029      CALL AX(A,N)
000030      DO 30 L=1,2
000031      30 YN1(L)=F(L)*EJ(K)+YN1(L)
000032      30 PN1(L)=(F(1)*F(L)+A(L))*EJ(K)+PN1(L)
000033      30 PN1(3)=(F(2)*F(2)+A(3))*EJ(K)+PN1(3)
000034      40 K=K+1
000035      J=NSD*NSD*RLANM2(1)*RLANM2(2)/FLOAT(H*H)
000036      40 YN1(1)=YN1(1)*0
000037      40 YN1(2)=YN1(2)*0
000038      40 PN1(1)=PN1(1)*0-YN1(1)*YN1(1)
000039      40 PN1(2)=PN1(2)*0-YN1(1)*YN1(2)
000040      40 PN1(3)=PN1(3)*0-YN1(2)*YN1(2)
000041      CALL XFM(PN1,RUN1,RLANM1)
000042      RETURN
000043      END

```

SUBPROGRAM LENGTH

000214

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLOCK NAMES AND LENGTHS

CM1 - 000003 CM2 - 000003

VARIABLE ASSIGNMENTS

```

A - 000176 D - 000213 F - 000174 I - 000207
J - 000211 K - 000213 L - 000212 M - 000000C01
HMAX - 000002C01 NSD - 000001C01 PN1 - 000020C02 RLANM1 - 000012C02

```

```

KFNL 1077
KFNL 1078
KFNL 1079

KFNL 1080
KFNL 1081
KFNL 1082
KFNL 1083
KFNL 1084
KFNL 1085
KFNL 1086
KFNL 1087
KFNL 1088
KFNL 1089
KFNL 1090
KFNL 1091
KFNL 1092
KFNL 1093
KFNL 1094
KFNL 1095
KFNL 1096
KFNL 1097
KFNL 1098
KFNL 1099
KFNL 1100
KFNL 1101
KFNL 1102
KFNL 1103
KFNL 1104
KFNL 1105
KFNL 1106
KFNL 1107
KFNL 1108
KFNL 1109
KFNL 1110
KFNL 1111
KFNL 1112
KFNL 1113
KFNL 1114
KFNL 1115
KFNL 1116

```

RLAHN2 - 000J02C02 RUN1 - 000J14C-2 RUN2 - 000004C02 S - 000203
 X - 000172 XP - 000201 YN1 - 000010C02 YN2 - 000000C02

START OF CONSTANTS

000150

START OF TEMPORARIES

000151

START OF INDIRECTS

000162

UNUSED COMPILER SPACE

036400

TT*****VARIABLE GIVEN CONFLICTING TYPES

000306


```

SUBROUTINE STATE(N)
  C
  GENERATES THE STATE VECTOR UPDATE
  DIMENSION SIG(4), SQU(2), AI(3)
  COMMON / CM1/ IT(3), A11, A12, A22, IT1(2), NP
  COMMON / CM3/ RT(3), A(3), IT(3), SQU(3)
  COMMON / CM4/ X(2), Z(2), U(2)
  DO 10 I=1,NP
    1) U(I)=GRND(0)
    UT=U(1)
    000012  U(1)=SQU(1)*U(1)
    000013  IF (NP.GT.1) U(2)=SQU(2)*UT+SQU(3)*U(2)
    000015  CALL SX(X,SIG,N)
    000023  CALL AX(SIG,A,N)
    000026  CALL MINV(A,AI,2)
    000032  A11=-.5*AI(1)
    000035  A12=-.5*AI(2)
    000037  A22=-.5*AI(3)
    000040  SQU(1)=SIG(1)*U(1)
    000042  IF (NP.GT.1) SQU(1)=SIG(1)+SIG(3)*U(2)
    000044  SQU(2)=SIG(2)*J(1)
    000052  IF (NP.GT.1) SQU(2)=SIG(2)+SIG(4)*U(2)
    000054  CALL X(X,X,N)
    000060  DO 20 I=1,2
    000063  2) X(I)=X(I)+SIG(I)
    000071  RETURN
    000073  END
    000074
  KFNLF 1117
  KFNLF 1118
  KFNLF 1119
  KFNLF 1120
  KFNLF 1121
  KFNLF 1122
  KFNLF 1123
  KFNLF 1124
  KFNLF 1125
  KFNLF 1126
  KFNLF 1127
  KFNLF 1128
  KFNLF 1129
  KFNLF 1130
  KFNLF 1131
  KFNLF 1132
  KFNLF 1133
  KFNLF 1134
  KFNLF 1135
  KFNLF 1136
  KFNLF 1137
  KFNLF 1138
  KFNLF 1139
  KFNLF 1140
  KFNLF 1141
  KFNLF 1142

```

```

SUBPROGRAM LENGTH
000123

```

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

```

BLOCK NAMES AND LENGTHS
CM1 - 00011 CM3 - 00014 CM4 - 00006

```

VARIABLE ASSIGNMENTS

```

A - 00002C02 AI - 000116 A11 - 00003C01 A12 - 00004C01
A22 - 00005C01 I - 000121 IT - 00006C01 IT1 - 00006C01
NP - 00011C01 RT - 00007C02 SIG - 00011
SQU - 00011C02 TT - 00006C02 U - 00009C03 UT - 00012
X - 00010C03 Z - 00010C03

```

START OF CONSTANTS

```
000077

```

START OF TEMPORARIES

```
000103

```

START OF INDIRECTS

```
000107

```

UNUSED COMMON ILLR SPACE

```
036600

```

```

000005 SUBROUTINE SX(X,S,N)
000005 DIMENSION X(1), S(1), T(4)
000005 COMMON / CM1/ NT(4), NP
000005 COMMON / CM3/ FL, SL, HL, ALF(4), ALS(4)
000005 LOGICAL FL, SL, HL
000005 IF (SL) GO TO 23
C NONLINEAR SIGMA COMPUTED HERE
C LINEAR IS SPECIAL CASE
000006 IF (NP.GY.1) GO TO 13
000006 S(1)=1.
000006 S(2)=1.
000006 RETURN
000006 1. S(1)=1.
000006 S(2)=0.
000006 S(3)=0.
000006 S(4)=1.
000006 RETURN
000006 2. S(1)=XLS(1)
000006 S(2)=XLS(2)
000006 S(3)=XLS(3)
000006 S(4)=XLS(4)
000006 RETURN
000006 END

```

SUBPROGRAM LENGTH:

000047

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

10 - 00015 23 - 000922

BLOCK NAMES AND LENGTHS

CM1 - 00011 CM8 - 000913

VARIABLE ASSIGNMENTS

FL - 00010, C02 HL - 000020, 2 NP - 00011, C01 NT - 000003, C02 XLS - 000007, C02

SL - 00001, C02 T - 000043 XLF -

START OF CONSTANTS

000033

START OF TEMPORARIES

- 000035

START OF INDIRECTS

000037

UNUSED COMPILER SPACE

037000

KFNLF 1143
KFNLF 1144
KFNLF 1145
KFNLF 1146
KFNLF 1147
KFNLF 1148
KFNLF 1149
KFNLF 1150
KFNLF 1151
KFNLF 1152
KFNLF 1153
KFNLF 1154
KFNLF 1155
KFNLF 1156
KFNLF 1157
KFNLF 1158
KFNLF 1159
KFNLF 1160
KFNLF 1161
KFNLF 1162
KFNLF 1163
KFNLF 1164
KFNLF 1165

```

000006      SUBROUTINE THUL T(A,B,C,V)
000010      DIMENSION A(1), 9(1), C(1)
000013      C(1)=A(1)*3(1)+A(2)*9(2)
000016      C(2)=A(2)*3(2)+A(3)*9(2)
000022      IF (N.EQ.1) RETURN
000026      C(3)=A(1)*3(3)+A(2)*3(4)
000030      C(4)=A(2)*3(3)+A(3)*8(4)
000031      RETURN
                                END

```

```

KFNL F      1166
KFNL F      1167
KFNL F      1168
KFNL F      1169
KFNL F      1170
KFNL F      1171
KFNL F      1172
KFNL F      1173
KFNL F      1174

```

```

SUBPROGRAM LENGTH
000046

```

```

FUNCTION ASSIGNMENTS

```

```

STATEMENT ASSIGNMENTS

```

```

BLOCK NAMES AND LENGTHS

```

```

VARIABLE ASSIGNMENTS

```

```

START OF CONSTANTS
000034

```

```

START OF TEMPORARIES
000035

```

```

START OF INDIRECTS
000037

```

```

UNUSED COMPILER SPACE
037100

```

KFNL F 1175
 KFNL F 1176
 KFNL F 1177
 KFNL F 1178
 KFNL F 1179
 KFNL F 1180
 KFNL F 1181
 KFNL F 1182
 KFNL F 1183
 KFNL F 1184
 KFNL F 1185
 KFNL F 1186
 KFNL F 1187
 KFNL F 1188
 KFNL F 1189
 KFNL F 1190
 KFNL F 1191
 KFNL F 1192
 KFNL F 1193
 KFNL F 1194
 KFNL F 1195

SUBROUTINE VMULT(K,L,M,N,A,B,C)
 DIMENSION A(1), B(1), C(1)

KL=K
 IF (L.EQ.1) KL=1

KM=K
 IF (M.EQ.1) KM=1

6 DO 200 II=1,L
 JK=0

LL=1
 DO 250 JJ=1,M

NN=JK
 MM=1

C(LL)=C.
 DO 100 KK=1,M

NN=NN+1
 C(LL)=C(LL)+A(M)*B(NN)

10 KK=MM+KL
 LL=LL+KL

20 JK=JK+KM
 RETURN
 END

SUBPROGRAM LENGTH
 000113

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS
 60 - 000120

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS					
C - 000100	II	-	000104	JJ	-
KK - 000112	KL	-	000102	KM	-
MM - 000120	NN	-	000110		

START OF CONSTANTS
 000070

START OF TEMPORARIES
 000071

START OF INDIRECTS
 000073

UNUSED COMPILER SPACE
 036700


```

000236 IC=-K+IXG2+1
000240 IF (IC.LT.1.OR.(G.GT.MMAX)) GO TO 40
000247 IC=(IC-1)*YMAX
000251 IS=-IXI(K)-IX2(K)+IXC1
000255 IF (IS.LT.1) IS=1
000260 IF (IS.GT.YMAX) IS=HMAX
000263 IS=IS+IC
000264 IL=IS+IX2(K)
000267 IF (IL.GT.IC+YMAX) IL=IC+HMAX
000272 GO TO 10
000273 40 CONTINUE
000301 C=C+M1
000302 S11=S11+M1*Y(1)*Y(1)
000304 S12=S12+M1*Y(1)*Y(2)
000306 S22=S22+M1*(Y(2)*Y(2))
000311 Y1=Y1+M1*Y(1)
000313 Y2=Y2+M1*Y(2)
000314 W(IJ)=W1
000316 50 IJ=IJ+1
000324 RETURN
000324 END

```

KFNLF 1250
 KFNLF 1251
 KFNLF 1252
 KFNLF 1253
 KFNLF 1254
 KFNLF 1255
 KFNLF 1256
 KFNLF 1257
 KFNLF 1258
 KFNLF 1259
 KFNLF 1260
 KFNLF 1261
 KFNLF 1262
 KFNLF 1263
 KFNLF 1264
 KFNLF 1265
 KFNLF 1266
 KFNLF 1267
 KFNLF 1268
 KFNLF 1269
 KFNLF 1270

SUBPROGRAM LENGTH

000370

FUNCTION ASSIGNMENTS

STATEMENT	ASSIGNMENTS	10	20	30	40	50
BLOCK NAMES AND LENGTHS						
CH1	000111 C12					
CH6	000107					
VARIABLE ASSIGNMENTS						
A11	000303CJ1 A12					
H2	000303 H3					
IJ	000352 IL					
IXC2	000356 IX1					
K	000361 M					
NPQ	000407CJ1 MSD					
RLAMN1	00012002 PLAMH2					
S11	00013004 S12					
M1	000357 XC					
XMN	000351 X41					
XZ	000101003 Y					
Y1	000101004 Z2					

START OF CONSTANTS

000327

START OF TEMPORARIES

000332

START OF INDIPECTS

000340

UNUSED COMPILER SPACE
039500

*****VARIABLE GIVEN CONFLICTING TYPES
000015,000015

```

000005      SUBROUTINE XFH(P,RU,RLAM)
000006      DIMENSION P(1), RU(1), RLAM(1)
000007      C      DETERMINES ROTATION MATRIX RU CORRESPONDING TO THE CORRELATION
000008      C      MATRIX P, STORED IN JMWER TRIANGULAR FORM. RLAM IS THE VECTOR
000009      C      OF SQUARE ROOTS OF EIGENVALUES.
000010      CALL EIGEN(P,RU,RLAM)
000011      RLAM(1)=SORT(RLAM(1))
000012      RLAM(2)=SORT(RLAM(2))
000013      RETURN
000014      END

```

KFNLF
KFNLF
KFNLF
KFNLF
KFNLF
KFNLF
KFNLF
KFNLF
KFNLF
KFNLF

1271
1272
1273
1274
1275
1276
1277
1278
1279
1280

```

SUBPROGRAM LENGTH
000037

```

```

FUNCTION ASSIGNMENTS

```

```

STATEMENT ASSIGNMENTS

```

```

BLOCK NAMES A'IC LENGTHS

```

```

VARIABLE ASSIGNMENTS

```

```

START OF CONSTANTS
000026

```

```

START OF TEMPORARIES
000027

```

```

START OF INDIRECTS
000035

```

```

UNUSED COMPILER SPACE
037100

```



```

000007 SUBROUTINE XP24(X, XU, XRL, XM, XP)
C DIMENSION X(1), XU(1), XRL(1), XM(1), XP(1)
C ROTATES THE VECTOR X INTO THE PRIMED VECTOR XP ACCORDING TO THE
C ROTATION MATRIX XU, THE SQUARE-ROOT EIGENVALUE VECTOR XRL, AND
C THE TRANSLATION VECTOR XM.
K=1
DO 10 I=1,2
XP(I)=0.
DO 10 J=1,2
XP(I)=XU(IK)*X(J)-XM(J)/XRL(I)+XP(I)
1: K=K+1
RETURN
END
000007 KENLF
000010 KENLF
000011 KENLF
000012 KENLF
000022 KENLF
000027 KENLF
000034 KENLF
000034 KENLF

```

```

1281 KENLF
1282 KENLF
1283 KENLF
1284 KENLF
1285 KENLF
1286 KENLF
1287 KENLF
1288 KENLF
1289 KENLF
1290 KENLF
1291 KENLF
1292 KENLF
1293 KENLF

```

```

SUBPROGRAM LENGTH
000055

```

```

FUNCTION ASSIGNMENTS

```

```

STATEMENT ASSIGNMENTS

```

```

BLOCK NAMES AND LENGTHS

```

```

VARIABLE ASSIGNMENTS

```

```

I - 000053 J - 000054 K - 000052

```

```

START OF CONSTANTS
000037

```

```

START OF TEMPORARIES
000040

```

```

START OF INDIRECTS
000045

```

```

UNUSED COMPILER SPACE
037100

```

```

000007      SUBROUTINE UNPR24(XP,XU,XRL,XM,X)
C          DIMENSION XP(1), XU(1), XRL(1), XM(1), Y(1)
C          ROTATES THE PRIMED VECTOR XP BACK TO THE UNPRIMED VERSION X,
C          ACCORDING TO THE ROTATION MATRIX XU, THE SQUARE-ROOT EIGENVALUE
C          VECTOR XRL, AND THE TRANSLATION VECTOR XM.
      DO 10 I=1,2
      K=I
      X(I)=XM(I)
      DO 10 J=1,2
      X(I)=X(I)+XU(K)*XP(J)*XRL(J)
      1, K=K+2
      RETURN
      END
000007
000011
000011
000013
000021
000025
000034
000035

```

```

KFNL F
KFNL F
KFNL F
KFNL F
KFNL F
KFNL F
KFNL F
KFNL F
KFNL F
KFNL F
KFNL F
KFNL F

```

```

1294
1295
1296
1297
1298
1299
1300
1301
1302
1303
1304
1305
1306

```

```

SUBPROGRAM LENGTH
000051

```

```

FUNCTION ASSIGNMENTS

```

```

STATEMENT ASSIGNMENTS

```

```

BLOCK NAMES AND LENGTHS

```

```

VARIABLE ASSIGNMENTS
I - 00046 J - 33353 K - 000347

```

```

START OF CONSTANTS
000040

```

```

START OF TEMPORARIES
000041

```

```

START OF INDIRECTS
000043

```

```

UNUSED COMPILER SPACE
037000

```

```

000014 SUBROUTINE YCNV(MC,N,M,EJ,JC,H1,H2,H3,IX1,IX2)
000015 INTEGE IX1(1),IX2(1)
000016 ***** YCNV
000017 C
000018 ***** YCNV
000019 ***** YCNV
000020 ***** YCNV
000021 ***** YCNV
000022 ***** YCNV
000023 ***** YCNV
000024 ***** YCNV
000025 ***** YCNV
000026 ***** YCNV
000027 ***** YCNV
000028 ***** YCNV
000029 ***** YCNV
000030 ***** YCNV
000031 ***** YCNV
000032 ***** YCNV
000033 ***** YCNV
000034 ***** YCNV
000035 ***** YCNV
000036 ***** YCNV
000037 ***** YCNV
000038 ***** YCNV
000039 ***** YCNV
000040 ***** YCNV
000041 ***** YCNV
000042 ***** YCNV
000043 ***** YCNV
000044 ***** YCNV
000045 ***** YCNV
000046 ***** YCNV
000047 ***** YCNV
000048 ***** YCNV
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000079 ***** YCNV
000080 ***** YCNV
000081 ***** YCNV
000082 ***** YCNV
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000088 ***** YCNV
000089 ***** YCNV
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000092 ***** YCNV
000093 ***** YCNV
000094 ***** YCNV
000095 ***** YCNV
000096 ***** YCNV
000097 ***** YCNV
000098 ***** YCNV
000099 ***** YCNV
000100 ***** YCNV

```

1361
1362
1363
1364
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1366
1367
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1370
1371
1372
1373
1374
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1380

KFNLF
KFNLF
KFNLF
KFNLF
KFNLF
KFNLF
KFNLF
KFNLF
KFNLF
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KFNLF
KFNLF
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KFNLF
KFNLF
KFNLF
KFNLF
KFNLF
KFNLF

IR=-K+IXC1+1
IF (IR.LT.1.OR.IR.GT.MMAX) GO TO 43
IS=IXC2-IX1(K)-IX2(K)
IF (IS.GT.MMAX) IS=MMAX
IF (IS.LT.1) IS=1
IS=(IS-1)*MMAX+IR
IL=IS+IX2(K)*MMAX
IF (IL.GT. IR+MMAX) IL=IR+MMAX
GO TO 13
40 CONTINUE
C=C+M1
Y1=Y1+M1*Y(1)
Y2=Y2+M1*Y(2)
S11=S11+M1*Y(1)*Y(1)
S12=S12+M1*(Y(1)*Y(2))
S22=S22+M1*Y(2)*Y(2)
M(IJ)=M1
50 IJ=IJ+1
RETURN
END

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SUBPROGRAM LENGTH

000402

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

10 - 000175 27 - 000227 31 - 000245 49 - 000303

BLOCK NAMES AND LENGTHS

CM1 - 000111 CM2 - 000129 CM4 - 00016 CM5 - 000006

VARIABLE ASSIGNMENTS

A11 - 000000 A12 - 000000 C - 000000 C04
H2 - 000001 I - 000001 IJ - 000001
IL - 000001 IS - 000001 IXC1 - 000001
IXC2 - 000001 IX2 - 000001 J - 000001
K - 000001 M - 000001 MMAX - 000001
N - 000001 NPQ - 000001 NSD - 000001
Q - 000001 RLANN1 - 000001 RLANN2 - 000001
RUN2 - 000001 S11 - 000001 S12 - 000001
Y1 - 000001 Y2 - 000001 XC - 000001
XK - 000001 XN - 000001 Y - 000001
XT - 000001 XZ - 000001 YN1 - 000001
YN2 - 000001 Y1 - 000001 Y2 - 000001
Z2 - 000001

START OF CONSTANTS

000136

START OF TEMPORARIES

000341

START OF INDIRECTS

000351

GENERALIZED TWO-DIMENSIONAL
DISCRETE NONLINEAR PREDICTOR

CAPTAIN KENNETH D. SENNE
FRANK J. SEILER RESEARCH LABORATORY
AIR FORCE SYSTEMS COMMAND
U.S. AIR FORCE ACADEMY, CO 80840
FEBRUARY 1971

MULTIMODAL RANGING TEST 29 MARCH 1971

PROBLEM PARAMETERS

MAXIMUM MONTE CARLO

SAMPLES PER PATH

1

100

GRID SIZE

NO. GRID SIGMAS

0. TRACKING ELLIPSE SIGMAS

M = 15

HSD = 6

NSDC = 16

NUMBER OF INPUTS

NUMBER OF OUTPUTS

INITIAL RANDOM NUMBER

IP = 2

HQ = 1

9152647325

PROCESS NOISE
COVARIANCE Q

.051 .025
.025 .051

MEASUREMENT NOISE
COVARIANCE R

.01J

LINEAR PARAMETERS

F

1.001 3.000
0.003 1.001

SIGMA

1.001 3.000
0.003 1.001

QWIKPLOT OUTPUT WITH 13 LEVELS
X LIMITS-HORIZONTAL- -6.070333E+70 6.070333E+00
Y LIMITS-VERTICAL- -6.070333E+70 6.070333E+00
Z LIMITS- 4.199141E-65 4.246628E-01

SAMPLE PATH INFORMATION FOR MONTE CARLO NUMBER 1

W VTH-1) ZTH-1) UTH-1) X(N) XHAT(W/N-1) SIGMA(W/N-1) XHAT(N/N-2) SIGMA(N/N-2) SQTLANU NPTS

0 7.75E+01 1.000E+01 1.000E+00 0. 1.00E+01 1.00E+00 0. 1.00E+00
 1.41E-01 0. 2.70E-01 0. 2.70E-01 5.19E-01 180

LINEARIZED CHECK- 3 ITERATION(7)...

1.00E+01 1.00E+00 0. 1.00E+01 1.00E+00 0. 2.70E-01

QUICKPLOT OUTPUT WITH 17 LEVELS
 X LIMITS-HORIZONTAL -6.00000E+00 5.00000E+00
 Y LIMITS-VERTICAL -6.00000E+00 6.00000E+00
 Z LIMITS 0. 2.4593272E-01

[illegible]

Additional Appendix B.

A Gaussian-Sum Program for the Passive Receiver Problem

This program was written for solving the "new problem" referred to in Appendix B of Chapter VI.

The program was compiled on one of the Kirtland AFB CDC 6600's using the RUN compiler. The random number generator is identical to the one used in the program of Additional Appendix A.

Programmer: D. L. Alspach

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```

PROGRAM KANLF(INPJ,OUTPJ,TAPE5=INPUT,TAPE6=OUTPUT)
DIMENSION X0(2)
COMMON/CSU/ CPRI(2,2,1500),DPRI(2,2,1500),BETA(1500),NPRI
COMMON/SSI/ AM(1,2),AP(1,2),BM(1,2,2),NM
COMMON/ONE/XSPRI(2,3),VSPRI(2,2,30)
COMMON/K2/EPK1(30),EPK2(30),EPS1(30),EPS2(30),AK(30),AS(30)
1,EPK(30),EPK2P(30),EPSIP(30),EPS2P(30),AKP(30),ASP(30)
1,NPK(30),XPR(30),XPR(30),XPR(30),ZPNT(30)
COMMON/3/3XPOST(2),VKPOST(2,2),VKPRI(2,2),ZK(1)
COMMON/4/4PHIT(2,2),H(1,2),PHIT(2,2),HT(2),Z(1),Q(2,2),R(1),NSTATE,
00012000
1 NMEAS=0, EPS=EP
COMMON/K8/1ST MEAS,TOFIM,TGSPRI,TGSPST,TKAL,TDIF,T0
1,KMAX,RHIN,RHOUT,BEFO,NRAND,KK
COMMON/K9/ A(3,3),AP(3,10),B(30,6),BP(30,6)
COMMON/S19/X(2),ZZZ,K,XSPST(2,30),VSPST(2,2,30)
NAMELIST/ONE/KK,N,NMAX,NRAND,KMAX,TMAX,NSKIP

READ(5,ONE)
97 FORMAT(1H--*KK=*,I4,* NMAX=*,I5,*NRAND=*,I5,*TMAX=*,E15.5)00020000
TMAX=THAX-8.
TST=L,STHEAS=L,STOF(1) )STGSPRI=0,STGSPST=0,STKAL=0,STDIF=0,SEP=0.00022000
WRITE(6,97) KK,N,NMAX,NRAND,TMAX
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C
DO 2, I=1,KMAX $DO 2, J=1,6
BP(I,J)=0.
2 R(I,J)=0.
CALL SECOND(TC)
WRITE(6,97) KK,N,NMAX,NRAND,TMAX
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C
DO 2, I=1,KMAX $DO 2, J=1,10
AP(I,J)=0.
2 A(I,J)=0.
IF (NSKIP .GT. 1) GO TO 1
IF (NSKIP .LE. 1) GO TO 1
t1=NSKIP*(3*KMAX+4)

C
1 N=N+1
NSTATE=2
NMEAS=1
RHIN=0.
RHAX=3.
NM=6
LPS=.1005
DELL=.0105
BET=1.
9ET=1.
X0(1)=3.
X0(2)=3.
XKPRI(1)=3.
XKPR(2)=3.
VKPRI(1,1)=1.
VKPRI(1,2)=0.
VKPRI(2,1)=0.
VKPRI(2,2)=1.
R(1)=.01
Q(1,1)=.1
Q(1,2)=.05

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000134 Q(2,1)=.05
000135 Q(2,2)=.1
000136 NPRI=1
000137 NM=5
000138 DELTA(1)=1.
000139 DO 200 I=1,2
000140   CPRI(I,1)=XCPRI(I)
000141   XSPRI(I,1)=XCPRI(I)
000142   DO 200 J=1,2
000143     JPRI(I,J,1)=VKPRI(I,J)
000144     VSPRI(I,J,1)=VKPRI(I,J)
000145   CONTINUE
000146   20 CONTINUE
000147   K=1
000148   PHI(1,1)=1.
000149   PHI(1,2)=0.
000150   PHI(2,1)=0.
000151   PHI(2,2)=1.
000152   CALL SECOND(T1)
000153   CALL FILGSV(KC,NM)
000154   CALL SECOND(T2) & TOTIX=TOTIM+T2-T1
000155   CALL SRGSKMAX)
000156
000157 C
000158   CALL SECOND(TC) & IF(TJ.GT.TMAX)GO TO 5
000159   IF(N.LT.NMAX) GO TO 1
000160   5 WRITE(6,101) T3
000161   WRITE(6,97) KK,NM,NMAX,VRAND,TMAX
000162   WRITE(6,89) TST,TMEAS,TOTIM,TGSPRI,TGSPST,T,KAL,TDIF,T0,EP
000163   89 FORMAT(1H-'9F11.5,')
000164   CALL OUTPGS(N,KMAX)
000165   STOP
000166   END
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PROGRAM LENGTH INCLUDING I/O BUFFERS

FUNCTION ASSIGNMENTS

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NH	-	00011602	NMAX	-	000373	NMEAS	-	000023C06	NPO	-	000550C04
NPR	-	00051604	NPRI	-	02404C01	NRAND	-	00015C07	NSKIP	-	000375
NSTATE	-	00022006	PHI	-	00000C06	PHIT	-	000006C06	Q	-	000015C06
R	-	00021006	RMAX	-	00012C07	RMIN	-	00001C07	YDIF	-	000006C07
TOSPRI	-	00000C07	YOSPRI	-	00000C07	YKAL	-	00000C07	YMAX	-	000374
TMEAS	-	00000C07	TOTIM	-	00000C07	TST	-	00000C07	T0	-	000007C07
T1	-	000401	T2	-	000402	VKPOST	-	00002C05	VKPRI	-	000010C05
VSPPOST	-	00000C11	VSPRI	-	00000C11	X	-	00000C11	XKPOST	-	000000C05
XKPRI	-	00000C05	XPNT	-	00000C05	XSPOST	-	00000C11	XSPRI	-	000000C03
X0	-	000371	Z	-	000014C06	ZK	-	000014C05	ZPNT	-	000740C04

START OF CONSTANTS

000322

START OF TEMPORARIES

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START OF INDIRECTS

000366

UNUSED COMPILER SPACE

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J00007 SUBROUTINE MEAS(Z,RMIN,RMAX,BET, R)
J00008 COMHON/GSI/ AM(1,2),A-PM(10),BM(10,2,2),NM
J00009 SR=SQRT(K)
J00010 TOT= .
J00011 DD=(RMAX-QMIN)/FLOAT(NM)
J00012 SZ=SIN(Z)
J00013 CZ=COS(Z)
J00014 SB=SIN(BET)
J00015 CB=COS(BET)
J00016 SZ2=SZ**2
J00017 CZ2=CZ**2
J00018 Q2I=1./(.7*DD)**2
J00019 R2I=1./SR**2
J00020 DO 30 L=1,NM
J00021 I=L
J00022 RN=FLOAT(L)*DD+R*IV
J00023 IF (NM.EQ. 1) RN=(RMAX-RMIN)*.5
J00024 A1=CZ/RN
J00025 A2=-SZ/RN
J00026 XH=SB+RN*SZ
J00027 YH=CB+RN*CB
J00028 A12=A1**2
J00029 A22=A2**2
J00030 A=A12*R2I+SZ2*Q2I
J00031 U=A22*R2I+CZ2*Q2I
J00032 G=-2.*((A12*XH+A1*A2*Y1)*R2I+(SZ2*SB+CB*SZ*CB+RN*SZ)*Q2I)
J00033 C=-2.*((A22*YH+A1*A2*X1)*R2I+(CZ2*CB+S8*SZ*CB+RN*CB)*Q2I)
J00034 C=2.*(R2I*A1*A2+R2I*SZ*CB)
J00035 BM(1,2,2)=A
J00036 BM(1,1,1)=3
J00037 BM(1,1,2)=C*.5
J00038 BM(1,2,1)=C*.5
J00039 UET=A*O-BM(1,1,2)**2
J00040 AM(1,2)=-.5*(O*B-E*C*.5)/OET
J00041 AM(1,1)=-.5*(-B*.5*C+A*E)/OET
J00042 ALPH(I)=1.
J00043 TOT=TOT+ALPH(I)
J00044 3 CONTINUE
J00045 TOTM1=1./TOT
J00046 DO 31 I=1,NM
J00047 31 ALPH(I)=ALPH(I)*TOTM1
J00048 RETURN
J00049 END

```

SUBPROGRAM LENGTH

J00400

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLOCK NAMES AND LENGTHS

GS1 - 000117

VARIABLE ASSIGNMENTS

A - JJ0371

ALPH - C J0024031 AM

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- 000363

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00132000
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AM2	=	000360	02	=	000359	082	=	000359	02	=	000352
SZ2	-	000355	0	-	000372	00	-	000347	DET	-	000376
E	-	000374	I	-	000361	L	-	000360	NM	-	000106C01
Q2I	-	000356	RN	-	000352	R2I	-	000357	S8	-	000352
SR	-	000345	SZ	-	000350	SZ2	-	000354	TOT	-	000346
TOTM	-	000377	XH	-	000369	YM	-	000366		-	

START OF CONSTANTS

000245

START OF TEMPORARIES

000254

START OF INDIRECTS

000337

JNJSEU COMPILER SPACE

J55200


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SUBROUTINE FILGSRV(XG,N4)
COMMON/ONE/XSPRI(2,30),VSPRI(2,2,30)
COMMON/CSO/ CPRI(2,150),OPRI(2,2,1500),BETA(1500),NPRI
DIMENSION X0(2)
COMMON/C15/ W(2),V(1)
COMMON/S19/X(2),ZZZ2,K,XSPST(2,30),VSPST(2,2,30)
COMMON/K5/XKPOST(2),VKPOST(2,2),XKPRI(2),VKPRI(2,2),ZK(1)
COMMON/ST/PHI(2,2),H(1,2),PHIT(2,2),HT(2),Z(1),Q(2,2),R(1),NSTATE
1 NMEAS,DELL,EPS,EP
COMMON/X8/TST,IMEAS,TOIFM,TGSPRI,TGSPST,TKAL,TOIF,T0
1,KMAX,RMIN,RMAX,BET,BETO,NRAND,KK
INITIALIZE
AA1=SQR(1.0(1,1))
AA2=Q(1,2)/AA1
AA3=SQR(1.0(2,2)-AA2*AA2)
C
JU 2,1 I=1,NSTATE
JO 2,1 J=1,NRAND
201 CALL RANSS(KK,W(1))
A1=SQR(VKPRI(1,1))
A2=VKPRI(1,2)/A1
A3=SQR(VKPRI(2,2)-A2*A2)
W(2)=W(1)*A2+W(2)*A3 $ W(1)=W(1)*A1
X(1)=X0(1)+W(1) $ X(2)=X0(2)+W(2)
C INITIAL STATE DETERMINED AND SET TO START MAJOR LOOP
1 CONTINUE
Z(1)=J
IF (K.GE.KMAX) GO TO 215
TAKE MEASUREMENT
S0=SIN(BET) $ C0=COS(BET)
Z(1)=ATAN2(X(2)-S0,X(1)-C0)
JO 2,3 I=1,NRAND
203 CALL RANSS(KK,V(1))
V(1)=V(1)*SQR(R(1)) $ Z(1)=Z(1)+V(1)
205 CONTINUE
ZZZ2=Z(1)
CALL BUFFGS
IF (NPRI*NN.GT.1500) GO TO 195
CALL SECOND(T1)
S2=SIN(Z(1)) $ C2=COS(Z(1))
RMI0=(XSPRI(2,K)-S0)*S2+(XSPRI(1,K)-C0)*C2
IF (RMI0.LE.0.) RMI0=0.
SIG=SQR(VSPRI(1,1,K)+VSPRI(2,2,K))
RMAX=RMI0+3.*SIG $ RMIN=RMI0-2.*SIG
IF (RMIN.LE.0.) RMIN=0.
ZZ=Z(1)
RR=R(1)
CALL MEAS(ZZ,RMIN,RMAX,BET,RR)
CALL SECOND(T2) $ IMEAS=IMEAS+T2-T1
CALL SECOND(T1)
CALL GSPOST
CALL SECOND(T2) $ IGSPST=IGSPST+T2-T1
CALL SECOND(T1)
CALL DROP2
CALL SECOND(T2) $ EP=EP+T2-T1
CALL SECOND(T1)
CALL COMBN3
CALL SECOND(T2) $ IOIF=IOIF+T2-T1

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JJ234 CALL SECOND(T1)
JJ236 CALL BUF2
JJ237 IF (K.GE. KMAX) GO TO 5
000235 CALL SECOND(T2) & TST=IST+T2-T1
      UPDATE TIME INDEX
      K=K+1
      BET=BET+BETD
      JO 3 I=1,NSTATE
      JO 3 J=1,NRAND
      JJ226 3 CALL RAMSS(KK,M(I))
      JJ227 M(2)=M(2)+AA2+M(2)*AA3 & M(1)=M(1)+AA1
      JJ276 M(2)=X(1) & X(1)=PHI(1,1)*X(1)+PHI(1,2)*X(2)+M(1)
      JJ230 X(2)=PHI(2,1)*TEMP2+PHI(2,2)*X(2)+M(2)
      JJ237 CALL SECOND(T1)
      JJ231 CALL GSPRI(PHI,Q)
      JJ233 CALL SECOND(T2) & TSPRI=TSPRI+T2-T1
      JJ220 GO TO 1
      JJ222 CONTINUE
      C
      C
      JJ222 RETURN
      JJ223 195 STOP
      JJ225 END

```

SUBPROGRAM LENGTH

000404

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

1 - J.0.56 5 - 000323 195 - 000324 205 - 000115

BLOCK NAMES AND LENGTHS

ONE - 000264 G50

K5 - 000 15 S7

K8 - 00027 K8

C15 - 000003 S19

K8 - 000017

000366

021450C02

000375

000012C06

000016C07

000022C06

000021C06

000401

000006C07

000001C07

000002C03

000074C01

000006C05

000014C05

000002C04

000002C04

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START OF CONSTANTS

JJ230

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SUBROUTINE BUFFGS
COMMON/ONE/XSPRI(2,3),VSPRI(2,2,30)
COMMON/CS4/CPOST(2,150),DPOST(2,2,1500),ALPHA(1500),NPOST
COMMON/CS0/CPRI(2,150),DPRI(2,2,1500),BETA(1500),NPRI
COMMON/K2/EPKI(30),EPK2(30),EPS1(30),EPS2(30),AK(30),AS(30)
1 EPK1(30)=EPK2(30),EPK2(30)=EPS1(30),EPS2(30)=AK(30),AK(30)=AS(30)
1,NP0(30),NPRI(30),XPNT(2,30),ZPNT(30)
COMMON/K5/XKPOST(2),VKPOST(2,2),XKPRI(2),VKPRI(2,2),ZK(1)
COMMON/S19/X(2),Z(2),XSPRI(2,30),VSPRI(2,2,30)
AA=1.E-10
XPNT(1,K)=X(1)*XPNT(2,K)=X(2)*ZPNT(K)=ZZZZ
NPO(K)=NPOST*NPRI(K)=NPRI
3 VSPOST(1,J,K)=.
DO 6 J=1,NPRI $ DO 60 J=1,2
XSPRI(J,K)=BETA(I)*CPRI(J,I)+XSPRI(J,K)
DO 61 L=1,2
6 VSPRI(J,L,K)=VSPRI(J,L,K)+BETA(I)*DPRI(J,L,I)+CPRI(J,I)*CPRI(L,I)
1)
6J CONTINUE
DO 62 J=1,2 $ DO 62 L=1,2
62 VSPRI(J,L,K)=VSPRI(J,L,K)-XSPRI(J,K)*XSPRI(L,K)
RETURN
ENTRY BUF2
DO 7 J=1,NPOST $ DO 70 J=1,2
XSPOST(J,K)=ALPHA(I)*CPOST(I,I)+XSPOST(J,K)
DO 75 L=1,2
75 VSPOST(J,L,K)=VSPOST(J,L,K)+ALPHA(I)*DPOST(J,L,I)+CPOST(J,I)*CPOST(L,I)
1(L,I)
7 CONTINUE
DO 72 J=1,2 $ DO 72 L=1,2
72 VSPOST(J,L,K)=VSPOST(J,L,K)-XSPOST(J,K)*XSPOST(L,K)
EPS1(K)=X(1)-XSPOST(1,K)
EPS1P(K)=X(1)-XSPRI(1,K)
EPS2(K)=X(2)-XSPOST(2,K)
EPS2P(K)=X(2)-XSPRI(2,K)
AS(K)=EPS1(K)*2*VSPOST(2,2,K)-2*VSPOST(1,2,K)*EPS1(K)*EPS2(K)+
1 EPS2(K)*2*VSPOST(1,1,K)
ASPK(K)=EPS1P(K)*2*VSPRI(2,2,K)-2*VSPRI(1,2,K)*EPS1P(K)*EPS2P(K)+
1 EPS2P(K)*2*VSPRI(1,1,K)
IF(VSPOST(1,1,K)+VSPOST(2,2,K)-VSPOST(1,2,K)*2
IF(DET.LI. AA*AS(K)) GO TO 71
IF(DET.LE. C)GO TO 71
AS(K)=AS(K)/DET
DET=VSPRI(1,1,K)*VSPRI(2,2,K)-VSPRI(1,2,K)*2
IF(DET.LI. AA*AS(K)) GO TO 71
IF(DET.LE. C)GO TO 71
ASPK(K)=ASPK(K)/DET
RETURN
71 WRITE(6,73) DET,AS(K),VSPOST(1,1,K),VSPOST(1,2,K),VSPOST(2,2,K)
WRITE(6,73) DET,ASPK(K),VSPRI(1,1,K),VSPRI(1,2,K),VSPRI(2,2,K)
WRITE(6,50)EPS1P(K),EPS2P(K),EPS1(K),EPS2(K),ASPK(K),AS(K),K
5. FORMAT(1H ,*CPKI=*,E12.5,2X,E12.5,2X,*EPS=*E12.5,2X,E12.5
,2X,*AK=*,E12.5,2X,*AS=*,E12.5,2X,*K=*,I4)
73 FORMAT(1H ,*ERROR IN VSPOST*,2X/, 7E12.4)
END

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SUBPROGRAM LENGTH
JJC474

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS	71		73	-	000416	
ONE	JJC474					
K5	JJC474					
BLOCK NAMES AND LENGTHS	GS4		GS0	-	024405	K2 - 000776
ONE	JJC474					
K5	JJC474					
VARIABLE ASSIGNMENTS						
AA	JJC474		AKP	-	000454004	ALPHA - 021450002
AS	JJC474		BETA	-	021450003	BUF2 - 000466
CP0ST	JJC474		DET	-	000473	OP0ST - 005570002
CPRI	JJC474		EPK1P	-	000264004	EPK2 - 000036004
EPK2P	JJC474		EPS1P	-	000360004	EPS2 - 000132004
EPS2P	JJC474		J	-	000471	K - 000030006
L	JJC474		NPOST	-	024404002	NPR - 000506004
NPRI	JJC474		VKPRI	-	000100005	VSP0ST - 000100006
VSPRI	JJC474		XKPRI	-	000000005	XKPRI - 000006005
XPNT	JJC474		XSPRI	-	000000001	ZK - 000014005
ZPNT	JJC474					

START OF CONSTANTS
JJC375

START OF TEMPORARILS
JJC423

START OF INDIRECTS
JJC437

UNUSED COMPILER SPACE
JJC444

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JJ-002      SUBROUTINE SRTGS(KMAX)
COMMON/K2/EPK1(3),EPK2(30),EPS1(30),EPS2(30),AK(30),AS(30)
1,EPK1P(30),EPK2P(30),EPS1P(30),EPS2P(30),AKP(30),ASP(30)
1,NP0(30),NPR(30),XPNT(30),ZPNT(30)
COMMON/K9/ A(3,10),AP(3,10),B(3,6),BP(30,6)
WR1(6,89)
3* FORMAT(1H-,'-----',/)
10 3 I=1,KMAX
4(I,3)=A(I,3)+EPS1(I)*A(I,4)+EPS2(I)
4(I,7)=A(I,7)+ABS(EPS1(I))*A(I,8)+ABS(EPS2(I))
4(I,10)=A(I,10)+AS(I)
5(I,4)=B(I,4)+EPS1(I)**2*B(I,6)+EPS2(I)**2
5(I,5)=B(I,5)+EPS1(I)*EPS2(I)
AP(I,3)=AP(I,3)+EPS1P(I)*AP(I,4)+EPS2P(I)
AP(I,7)=AP(I,7)+ABS(EPS1P(I))*AP(I,8)+ABS(EPS2P(I))
AP(I,10)=AP(I,10)+ASP(I)
3P(I,4)=BP(I,4)+EPS1P(I)**2*3P(I,6)+EPS2P(I)**2
4P(I,5)=4P(I,5)+EPS1P(I)*EPS2P(I)
5 CONTINUE
WR1(6,93)(I,EPK1P(I),EPS2P(I),ASP(I),EPS1(I),EPS2(I),AS(I),NPR(I))
1),XPNT(1,I),XPNT(2,I),ZPNT(I),I=1,KMAX)
9* FORMAT(1H ,13,0F10.5,I,3F10.5)
*EJEN
END
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SUBPROGRAM LENGTH
JJ-231

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS
99 - JJ-239 93 - JJ-141

BLOCK NAMES AND LENGTHS
K2 - JJ-276 K9 - JJ-160

VARIABLE ASSIGNMENTS

A - JJ-2 C-2 AK - JJ-17001 AKP - 000454001 AP - 000454002
AS - JJ-20001 ASP - JJ-12001 B - 001130002 BP - 001414002
EPK1 - JJ-2 C-1 EPK1P - JJ-26001 EPK2 - 000036001 EPK2P - 000322001
EPS1 - JJ-7001 EPS1P - JJ-35001 EPS2 - 000132001 EPS2P - 000416001
I - JJ-23 NP0 - 000506001 NPR - 000606001 XPNT - 000544001
ZPNT - JJ-2 C-1

START OF CONSTANTS

JJ-133

START OF TEMPORARIES

JJ-145

START OF INDIRECTS

JJ-166

UNUSED COMPILER SPACE

JJ-548

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SUBROUTINE GPOST
COMMON/GS1/ AM(10,2),ALP(10),BM(10,2,2),NM
COMMON/USU/ CPRI(2,1500),DPOST(2,2,1500),BETA(1500),NPRI
COMMON/CS4/ CPST(2,1500),CPST(2,2,1500),ALPHA(1500),NPST
DIMENSION B(10,2,2),S(2,2),D(2),PM1(2,2),WM(2,2),QM1(2,2)
TC=0., $ NPST=NM+VPMI $L=0 $ DO 1 I=1,NM
  DET=BM(I,1,$)*M(I,2,2)-BM(I,1,2)**2 $ DETM1=1./DET
  S(1,1)=BM(I,2,2)*D(1,1) $ B(1,2,2)=BM(I,1,1)*DETM1
  S(1,2,2)=-BM(I,1,2)*DETM1 $ B(1,2,1)=B(1,1,2)
  CONTINUE
  DO 3 J=1,NPRI
    DET=CPRI(1,1,J)*CPRI(2,2,J)-OPRI(1,2,J)**2 $ DETM1=1./DET
    PM1(1,1)=DET*OPRI(2,2,J) $PM1(2,2)=DETM1*OPRI(1,1,J)
    PM1(1,2)=-DETM1*CPRI(1,2,J) $PM1(2,1)=PM1(1,2)
    DO 3 I=1,NM
      S(1,1)=1./NM
      S(1,2)=PM1(1,2)+3M(I,1,L2)+3M(I,1,L2)
      DETQ=QM1(1,1)*QM1(2,2)-QM1(1,2)**2 $ DETQM1=1./DETQ
      PM1(1,1,L)=QM1(2,2)*DETM1 $DPOST(2,2,L)=QM1(1,1)*DETM1
      PM1(1,2,L)=-QM1(1,1,2)*DETM1 $DPOST(2,1,L)=DPOST(1,2,L)
      DO 3 L1=1,2 $DO3L2=1,2$M(L1,L2)=0.
      DO 3 L3=1,2
        WK(L1,L2)=WK(L1,2)*DPOST(L1,L3,L)*8M(I,L2,L3)
      DO 4 L1=1,2
        U(L1)=AM(I,L1)-CPRI(L1,J)
      DO 5 L1=1,2
        UPM1(L1,L)=CPRI(L1,J)+WK(L1,1)*D(1)+WK(L1,2)*D(2)
        S(1,1)=3(I,2,2)+CPRI(2,2,J)
        S(2,2)=B(1,1,2)+CPRI(1,1,J)
        S(1,2)=-B(1,1,2)-OPRI(1,2,J)
        DETMI=1./S(1,1)*S(2,2)-S(1,2)**2
        S(1,1)=S(1,1)*DETM1 $ S(2,2)=S(2,2)*DETM1 $ S(1,2)=S(1,2)*DETM1
        W(1,1)=S(1,1)*S(1,1)+D(2,2)**2*S(2,2)*S(1,2)
        ALPHA(L1)=9*ETA(1)*ALP(I)*EXP(-.5*W(1,1))
        TOT=TOT+ALPHA(L1)
      TOTMI=1./TOT
    DO 5 I=1,NPST
      ALPHA(I)=ALP(I)*TOTMI
    PRINTN
  END

```

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SUBPROGRAM LEIGH

435

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLACK NAMES AND LENGTHS

GS1	- JUL 1	GS4	-	024405
-----	---------	-----	---	--------

VARIABLE ASSIGNMENTS

ALPHA	215 C13 ALPH	-	UJ24C11 AM	-	000000C01 8	-	000326
BETA	215 C2	-	UJ36C11 CPOST	-	000000C03 C8R1	-	000000C02
DELTA	215 C2	-	UJ23 DETM1	-	000424 DETQ	-	000000C00

GEIQM1 -	JJ431	DPOST	-	005670C02	I	-	000422
J	JJ423	L	-	00421	L1	-	000427
L3	JJ432	NM	-	0010C01	NPOST	-	024404C02
PM1	JJ414	QM1	-	00414	S	-	000420
TOTM1	JJ434	WK	-	00413	MT1	-	000433

START OF CONSTANTS

JJ262

START OF TEMPORARIES

JJ266

START OF INDIRECTS

JJ330

UNUSED COMPILER SPACE

J55700

```

SUBROUTINE GSPRI(Phi,Q)
  DIMENSION PHI(2,2),Q(2,2),T(2,2),C(2)
  COMMON/GS0/ CPRI(2,1500),OPRI(2,2,1500)-BETA(1500),NPRI
  ONE STAGE PRELICTED DENSITY *** ALSO INITIAL STATE PDF
  DIMENSION TE(2,2)
  C
  A POSTERIOR: DENSITY
  C
  COMMON/GS4/CPOST(2,1500),JPOST(2,2,1500),ACPHA(1500),NPOST
  SYSTEM FORMULATION AND SET UP
  NPRI=NPOST
  JO 3 I=1,NPRI
  BETA(I)=ALPHA(I)
  JO 4 J=1,2 $ D3 4 L=1,2
  4 TEMP4(J,L)=DPOST(J,L,I)
  T(1,1)=PHI(1,1)*2*TEMP4 (1,1)+2*PHI(1,1)*TEMP4 (1,20034000
  2)+PHI(1,2)*2*TEMP4 (2,2) 00341000
  T(2,2)=PHI(2,1)*2*TEMP4 (1,1)+2*PHI(2,2)*TEMP4 (1,20034000
  2)+PHI(2,2)*2*TEMP4 (2,2) 00342000
  T(1,2)=PHI(1,1)*PHI(2,1)*TEMP4 (1,1)+PHI(2,2)*TEMP4 (2,1))+
  1*PHI(2,2)*PHI(1,2)*TEMP4 (2,2)+PHI(2,1)*PHI(1,2)*TEMP4 (1,2)
  T(2,1)=T(1,2)
  JO 5 J=1,2$D3 5 L=1,2
  5 OPRI(J,L,I)=T(J,L)+Q(J,L)
  2 C(J)=CPOST(J,I)
  C(1)=PHI(1,1)*C(1)+PHI(1,2)*C(2)
  C(2)=PHI(1,2)*CPOST(1,1)+PHI(2,2)*C(2)
  JO 5 J=1,2
  3 CPRI(J,I)=C(J)
  CONTINUE
  RETURN
END

```

SUBPROGRAM LENGTH
J05212

=FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

1 - J0J142

BLOCK NAMES AND LENGTHS
GS. - J2465 GS4 - J24405

VARIABLE ASSIGNMENTS

ALPHA - J21-5 C12 BETA - J2145J0C11 C - 000201 CPOST - 000000C02
CPRI - J00-5 C11 UPOST - J0567J0C01 I - 000207
J - J00-1 L - C0011 NPOST - 024404C02 NPRI - 024404C01
T - J00-75 TEMP4 - J0J203

START OF CONSTANTS

J0C145

START OF TEMPORARILS

J0C147

START OF INDIRECTS

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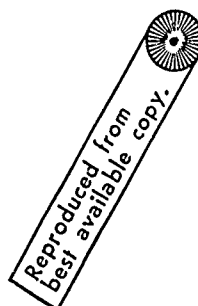

```

SUBROUTINE COMB3
  DIMENSION CMN(2,2), AMN(2)
  COMMON/CS4/CP(2,1),JC(2,2,1500),ALP(1500),NP
  COMMON/CS5/MA(2),TEMP(1,2),TEMP1(1),TEMP2(2),TEMP3(2,2),C(2),A(1)
  1,AT(2),AA(2)
  COMMON/ST/PHI(2,2),H(1,2),PHI(2,2),HT(2),Z(1),Q(2,2),R(1),NSTATE,00396000
  1 NMEAS, JELL, EPS, EP
  CALL SECOND(T1)
  N=NSTATE $ M=NMEAS
  ZZ=Z(1)
  DO 11 I=1,NP
    AMN(2)=CP(2,I) $L=I
    DO 12 J=1,NP
      IF (CP(2,J)-AMN(2))13,12,12
    13 CONTINUE
    AMN(2)=CP(2,J) $ L=J
  14 CONTINUE
  3MN=ALP(L)
  ALP(L)=ALP(I) $ALP(I)=3MN
  AMN(1)=CP(1,L)
  DO 14 K1=1,2
    CP(K1,L)=CP(K1,I) $CP(K1,I)=AMN(K1)
  NO 14 K2=1,2
  CMN(K1,K2)=DP(K1,K2,L)
  1P(K1,K2,L)=DP(K1,K2,I)
  1P(K1,K2,I)=CMN(K1,K2)
  14 CONTINUE
  15 CONTINUE
  16 CONTINUE
  DO 16 K=1,NP
    WK=ALP(K)
    LNC= $ L1=1
    IF (L1.EQ. K) GO TO 3
    J=K-L1
    Z(I)=0.
    WJ=ALP(J)
    IF (WJ.EQ.0.)GO TO 41
    WM1=1./ (WK+WJ)
    A(1)=WJ*WK*WM1
    JELLS=(JELL/A(1))*2
    DO 2 I=1,N
      Z(I)=Z(I)+(CP(I,I)-CP(I,J))*2
    IF (Z(1).GT. DILLS) GO TO 42
    DO 2 L=1,N
      Z(L)=Z(L)+ABS(JP(L,I)-JP(L,I,K))
    IF (Z(1).GT. DILLS) GO TO 42
  2 CONTINUE
  DO 3 I=1,N
    CP(I,J)=(CP(I,I)*WJ+WK*CP(I,K))*WM1
    DO 3 L=1,N
      3P(I,L,J)=(CP(I,L,J)*WJ+WK*3P(I,L,K)+A(1))*(CP(I,J)-CP(I,K))*CP(L,
    1J)-CP(L,K))*WM1
  6 CONTINUE
  ALP(K)=0.
  3P(K)=WK+WJ
  GO TO 44
  7 1=M+1
  IF (K.EQ.M)GO TO 40

```

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00461000
00462000
00463000
00464000



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JJ233      ALP(M)=ALP(K)SALP(K)=0.
JJ236      DO 5 I=1,N
JJ237      CP(I,M)=CP(I,K)
JJ243      DO 5 L=1,N
JJ257      DP(I,L,M)=DP(I,L,K)
JJ258      5 CONTINUE
JJ262      GO TO 40
JJ262      42 LNC=LNC+1
JJ264      IF (LNC.GT.2) L1=K-1
JJ267      41 L1=L1+1
JJ271      GO TO 43
JJ271      4 CONTINUE
JJ274      NP=N
JJ275      Z(I)=ZZ
JJ276      RETURN
JJ277      END

```

SUBPROGRAM LENGTH

JJ240

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

2 - JJ215 3
40 - JJ270 41

BLOCK NAMES AND LENGTHS

354 - JJ240 56

VARIABLE ASSIGNMENTS

A - JJ240 2 AA
AT - JJ240 2 BMN
DELL - JJ240 3 DCLLS
I - JJ240 3 HI
K - JJ274 K1
LNC - JJ273 L1
NMEAS - JJ273 NP
PHIT - JJ273 O
TEMP1 - JJ273 2 TEMP2
WA - JJ273 2 WJ
Z - JJ273 2 Z

START OF CONSTANTS

JJ302

START OF TEMPORARIES

JJ304

START OF INDIRECTS

JJ323

UNUSED COMPILER SPACE

JJ370

12 - 00025 13 - 00021
42 - 00026 43 - 00010

57 - 00027

00022JC02 ALP - 021450C01 AMN - 000355
L0360 C - 000013C02 CP - 000000C01
JJ377 OMN - 000351 DP - 005570C01
00012C03 I - 000363 J - 000365
JJ367 K2 - 000379 L - 000364
JJ374 M - 000361 N - 000360
0240-031 NSTATE - 00022C03 PHI - 000000C03
JJ15C03 R - 000021C03 TEMP - 000002C02
00005C02 TEMP4 - 000037C02 T1 - 000357
JJ775 WK - 000372 MM1 - 000376
JJ362

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```

      FUNCTION ARMX(X,N)
      DIMENSION X(N)
      ARMX=X(1)
      DO 1 J=2,N
      IF(X(J)-ARMX)1,1,2
      2 ARMX=X(J)
      1 CONTINUE
      RETURN
      END
      JJ004
      JJ004
      JJ005
      JJ0011
      JJ0013
      JJ0014
      JJ0015
      JJ0017

```

SUBPROGRAM LENGTH
JJ0031

FUNCTION ASSIGNMENTS

--STATEMENT ASSIGNMENTS
1 - JJ0015 2 - JJ0014

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS
--ARMX - JJ0027 J - JJ0033

START OF CONSTANTS
JJ0022

START OF TEMPORARIES
JJ0023

START OF INDIRECTS
JJ0025

UNUSED COMPILER SPACE
JJ0026

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SUBROUTINE RANSS(KK,H)
COMMON / CM6/ PI, THOP1
C01404 / CHIL/ ISF, IR(2), X(2)
IA1A (IT=1)
IR(2)=KK
NN=0
IF (IT.NE.1) GO TO 13
IT=0
PI=4.*ATAN(1.)
THOP1=2.*PI
IST=KK
IR(1)=0
NN=1
1 H=GRN(1)(NN)
KK=IR(2)
RETJRN
END

```

SUBPROGRAM LENGTH

J00044

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

1. - 00023

BLOCK NAMES AND LENGTHS

CH6 - 00002 CH10 - 00005

VARIABLE ASSIGNMENTS

IR - 00002 IST - 00002
PI - 00001 THOP1 - 00002
X - 00002

START OF CONSTANTS

00002

START OF TEMPORARIES

00006

START OF INDIRECTS

00002

UNUSED COMPILER SPACE

00002

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```

      FUNCTION GRN1 (NM)
      COMMON / CMO / PI, T40PI
      DIMENSION / C41 / IST, IR(2), X(2)
      DATA (IM=543755815887), (JM=2513271)
      C=12*(1)
      IF (NM.EQ.1) GO TO 1,
      GO TO (2), (4), J
      1 X(2)=IST
      2 X(1)=2
      3 X(1)=FLOAT(12*(2))/FLOAT(IM)
      X1=SQRT(ABS(-2.*ALOG(X(1))))
      X(2)=1*CHI(X(2))
      X(1)=X1*SIN(X(2))
      X(1)=X1*COS(X(2))
      GRN1=X(1)
      RETURN
      4 IR(1)=1
      GRN1=X(2)
      RETURN
      END

```

SUBPROGRAM LEIGH
...120

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

1J - J J 15 2- - 1J0015 40 - 000057

BLOCK NAMES AND LENGTHS

NAME - J J 15 2- - 1J0015

VARIABLE ASSIGNMENTS

GRN1 - J J 15 2- - 1J0015
IST - J J 15 2- - 1J0015
X - J J 15 2- - 1J0015

START OF CONSTANTS

J...04

START OF TEMPORARIES

J...056

START OF INDIRECTS

J...111

UNUSED COMPILER SPACE

057000

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SUBROUTINE OUTPGS(N,KMAX)
COMMON/K9/ A(3,10),AP(33,10),B(33,6),BP(30,6)
DO 10 I=1,10
  WRITE(0,99) I
  *****
  95 FORMAT(1H-,1JF11.2)
  FFF=C(4)
  DO 7 I=1,1-
    L(I)=C(1)/FFF
  WRITE(0,99) C
  99 FFF=1A(1H-,1JF11.7)
  1 CONTINUE
  PRINT 95
  *****
  95 FORMAT(1H1)
  FFF=1./FLOAT(N)
  DO 7 I=1,KMAX 5 DO 4 J=1,13
    AP(I,J)=AP(I,J)*FNM1
  4 A(I,J)=A(I,J)*FNM1
  DO 7 I=1,KMAX500 41 J=1,6
    BP(I,J)=BP(I,J)*FNM1
  41 B(I,J)=B(I,J)*FNM1
  WRITE(0,93) (A(I,3),A(I,4),B(I,4),B(I,5),B(I,6),A(I,7),A(I,8),
    1 A(I,1),I=1,KMAX)
  C
  C
  9. FORMAT(1H,2F11.7,2X,3=11.7,5X,2F11.7,2X,F11.7)
  10. FORMAT(8F11.6)
  PRINT 95
  1.1. FORMAT(1H-,*APRIRI*,/)
  WRITE(0,1003)
  WRITE(0,93) (AP(I,3),AP(I,4),BP(I,4),BP(I,5),BP(I,6),AP(I,7),
    1AP(I,8),AP(I,1),I=1,KMAX)
  C
  C
  5J(2:4) -ETURN
  1.1.2.5 -NU

```

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SUBPROGRAM LENGTH
J.253

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS	95	98	90213
10 - J 52 93	- J 226	- 000223	- 90213
99 - J 17 1.3	- J 234	- 000236	

BLOCK NAMES AND LENGTHS
K9 - J 1/6 R10 - 00012

VARIABLE ASSIGNMENTS

VARIABLE	3	00139C01 BP	00144C01
A - J 1 J.1 AP	- J 454J11 3	- 00139C01 BP	- 00144C01
C - J 1 C.2 FFF	- J 247 FNM1	- 000251 I	- 001250
J - J 253			

START OF CONSTANTS

JJL21,

START OF TEMPORALS

JJ241

START OF INDIRECTS

JJ245

JNJSED COMPIL & SPACE

J5E2J,

VP*****NO PAT. TJ THIS STATEMENT

UJJJJ5

00578000
00579000
00580000
00581000

SUBROUTINE SECOND(T)
T=...
RETURN
END

SUBPROGRAM LENGTH
000011

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS

START OF CONSTANTS

JJ006

START OF TEMPORARIES

JJ007

START OF INDIRECTS

JJ011

UNUSED COMPILER SPACE

057200

KK= 51526+87320N= J NMAX= 100NRAND= 1THAX 00000E+01
IJ=

1	15197	51644	63953	76019	113585	126342	1	315197	34644	81210
2	94347	152777	198831	38987	77459	164335	6	333525	42837	69240
3	37335	114051	332706	17058	71078	202732	17	331873	457429	83475
4	3213	38481	51056	19660	23056	79088	35	318027	424433	85345
5	32233	3342	174059	43750	77662	98889	47	305055	444718	120653
6	34744	59755	7438	17091	57945	72456	46	296449	436811	112711
7	49749	57851	47332	39526	73332	72285	39	329107	436718	86540
8	69593	131733	192129	54902	60948	52273	41	359174	495119	94448
9	17977	91737	161559	100205	80433	176658	54	412249	525908	79750
10	91439	46579	12142	88399	30179	249081	51	39383	486055	96542
11	106340	69751	185398	83379	57949	155351	50	411624	525627	100103
12	101469	2375	208127	21600	43194	25177	43	429514	520053	80437
13	66558	2827	157577	75892	109814	272146	41	341356	474032	91400
14	9667	6841	171211	96595	97113	180663	47	320542	423005	88408
15	108257	12958	19486	97123	145379	314467	53	308980	417160	93784
16	166255	146085	46792	175876	150702	578900	50	239847	416454	83599
17	101325	99859	573059	80783	60188	149017	43	234398	467296	133625
18	79444	34413	133578	25593	75022	116886	45	235337	453072	136059
19	1406	29226	13299	4668	25616	29565	37	247164	498868	133405
20	23043	32966	21111	23805	37986	29101	35	219452	491512	113350
21	37309	47697	40226	26622	256094	4135426	39	205858	481807	00000

Additional Appendix C.

A Gauss-Hermite Program for Implementing the Two-Dimensional Phase Demodulator

This program was written to test the Gauss-Hermite method. The numerical results of this program are given in Appendix B of Chapter VII.

The program was compiled on one of the Kirtland AFB CDC 6600's using the FTN compiler. The code-optimization algorithm was employed. The random number generator is identical to the one described in Chapter IV.

Programmer: C. Hecht

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1

PAGE

CDC 660C FIN V3.0-P213 OPT=2 03/19/72 00.54.44.

PROGRAM MAIN

```

PROGRAM MAIN(INPUT,OUTPUT,TAPES=INPUT,TAPE6=OUTPUT)
COMMON/NAME6/NUM1:NO1,NO2
COMMON/NAME8/A11,A22,Q22,RR11,DELT,JSEED
COMMON/NAME9/Y1EST,Y2EST
COMMON /RN/ NZZ(3), XNZ(2)
COMMON /GN/ DZZ(3), JGAUSS, XZZ(2)
JGAUSS=J

```

```

100 CONTINUE
101 READ(5,651)Y1EST,Y2EST,P110,DELT,Q22C,NUM1,NO1,NO2,JSEED
IF(EOF(5))102,101

```

```

102 CONTINUE
P110=P110**2
QQ=Q22C**(.25)
RR=(P110/(SQRT(2.0)*QQ))**(.40/3.0)
FTC=SQRT(2.0)* RR**(.25) /QQ
DELT=DELT*FTC
Q22=Q22C*DELT
R11=RR*DELT
P220=P110*SQRT(Q22C/RR)
A11= P110
A22= P220

```

```

WRITE(6,652) P110,FTC,DELT,Q22,R11

```

```

WRITE(6,657) A11,A22

```

```

657 FORMAT(*,30X,*VARIANCE OF INIT. POS. AND VEL.*,1P2E15.6)

```

```

WRITE(6,653) Q22C,RR

```

```

653 FORMAT(*Q*,*CONTINUOUS DRIVING VARIANCE =*,1P2E14.6/

```

```

1 * CONTINUOUS NOISE VARIANCE =*,1P2E14.6)

```

```

651 FORMAT(5E10.4,3I5,110)

```

```

652 FORMAT(*1*,*EQUILIBRIUM CONTINUOUS POSITION VARIANCE =*,1P2E14.6/
1 * DELTA =*
1 * DISCRETE DRIVING VARIANCE =*
1 * DISCRETE NOISE VARIANCE =*
CALL INITIAL
CALL CYCLE

```

```

IF(N02.LT.511) GO TO 100

```

```

102 CONTINUE

```

```

STOP

```

```

END

```

PROGRAM MAIN

SYMBOLIC REFERENCE MAP

ENTRY POINTS DEF LINE REFERENCES

4052 MAIN

VARIABLES SN TYPE RELOCATION

0 A11 REAL NAME8

1 A22 REAL NAME8

312 BINOM REAL NAME

4261 DELF REAL NAME

4 DELT REAL NAME8

0 02271 REAL GN

4265 FTC REAL NAME

50 GAMA REAL NAME

171 IM REAL GN

1 IGAD11 INTEGER NAME8

3 JSEFO INTEGER NAME6

1 NOX INTEGER NAME6

2 NO2 INTEGER NAME6

0 NU41 INTEGER NAME6

6 N127 INTEGER NAME6

4260 P110 REAL NAME6

4266 P223 REAL NAME6

4263 QQ REAL NAME6

2 Q21 REAL NAME6

4262 Q220 REAL NAME6

4264 RR REAL NAME6

3 R11 REAL NAME6

0 T REAL NAME6

24 W REAL NAME6

3 XNZ REAL NAME6

2 XZZ REAL NAME6

6 Y1EST REAL NAME6

1 Y2EST REAL NAME6

FILE NAMES MODE

0 INPUT

2022 OUTPUT

0 TAPE5

2022 TAPE6

EXTERNALS TYPE PGS REFERENCES

CYCLC 0 36

EOF 1 11

INITIAL 0 35

SQRT 1 LIBRARY 15

16

20

24

26

26

26

26

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26

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26

26

26

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26

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26

STATEMENT LABELS DEF LINE REFERENCES

4054 100 9 37

0 101 12 11

4174 102 38 11

4222 651 29 10

4225 652 30 23

21 24

22 24

23 10

19 19

16 16

23 23

8 8

10 10

10 10

37 37

10 10

21 21

20 20

13 13

15 15

20 20

20 20

14 14

16 16

23 23

18 18

20 20

18 18

20 20

20 20

23 23

19 19

10 10

10 10

10 10

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CDC 6600 FTN V3.0-P213 OPT=2 03/19/72 00.54.44.

PROGRAM MAIN

STATEMENT LABELS	DEF LINE	REFERENCES
4210 653 FMT	27	26
4202 657 FMT	25	24

COMMON BLOCKS NAME LENGTH

NAME	491		
NAME6	3	20 NT (20)	40 GAMA (81)
NAME8	6	202 BINOM (289)	2 NO2 (1)
NAME9	2	1 NO1 (1)	2 Q22 (1)
RN	5	1 A22 (1)	5 JSEED (1)
CN	4	4 DELT (1)	
		1 Y2EST (1)	
		3 XNZ (2)	
		1 JGAUSS (1)	2 XZZZ (2)

STATISTICS

PROGRAM LENGTH	2258
BUFFER LENGTH	40448
COMMON LENGTH	7778

MEMBERS - BIAS NAME(LENGTH)

0 I (20)
121 HM (81)
1 NUM1 (1)
0 A11 (1)
3 R11 (1)
0 Y1EST (1)
0 NZZZ (3)
1 DZZZ1 (1)

SUBROUTINE INITIAL

1

PAGE

COC 6600 FTN V3.0-P213 OPT=2 03/19/72 00.54.44.

```

SUBROUTINE INITIAL
COMMON/NAME/I(20),WT(20),JAMA(9,3),HM(9,9),BINOM(17,17)
COMMON/NAME6/NUM1,N01,N02
CALL HERMIT(NUM1,I,WT)
CALL COEF(N01,HM)
CALL CHBTRL(N01,BINOM)
CALL GAM22(N01,GAMA)
RETURN
END
    
```

5

SUBROUTINE INITIAL

SYMBOLIC REFERENCE MAP

ENTRY POINTS	DEF LINE	REFERENCES
1 INITIAL	1	8

VARIABLES	SN	TYPE	ARRAY	NAME	RELOCATION	REFS
312 BINOM		REAL	ARRAY	NAME		2
50 GAMA		REAL	ARRAY	NAME		2
171 HM		REAL	ARRAY	NAME		3
1 NO1		INTEGER		NAME6	6	3
2 NO2		INTEGER		NAME6	7	3
0 NUM1		INTEGER		NAME6		3
0 T		REAL	ARRAY	NAME		2
24 WT		REAL	ARRAY	NAME		2

EXTERNALS	TYPE	ARGS	REFERENCES
CMTRL		2	6
COEF		2	5
GAM22		2	7
HERMIT		3	4

COMMON BLOCKS	LENGTH	MEMBERS - BIAS NAME(LENGTH)
NAME	491	0 T (20)
NAME6	3	121 HM (81)
		8 NUM1 (1)

20 WT (20)	40 GAMA (81)
202 BINOM (289)	2 NO2 (1)
1 NO1 (1)	

STATISTICS

PROGRAM LENGTH	308	24
COMMON LENGTH	7566	494

SUBROUTINE HERMIT

PAGE 1

CDC 6600 FTN V3.0-P213 OPT=2 03/19/72 00.54.44.

```

SUBROUTINE HERMIT (NO,I,WT)
  DIMENSION T(20),WT(20)
  DIMENSION T20(10),WT20(10)
  DIMENSION T10(5),WT10(5)
  DATA T10/.3423013272,1.0366108297,1.7566836492,2.5327316742,
    1 3.4361591188/
  DATA WT10/.1086263373E-12,4.0139611J8E-1,3.3874394455E-2,
    1 1.34335457487E-3,7.6404328552E-5/
  DATA T20/.24534076,0.73747373,1.23407621,1.73853771,2.25497400,
    X 2.78880606,3.34785657,3.94476404,4.60368245,5.38748089/
  DATA WT20/.422243659E-1,2.866755,5E-1,1.09017246E-1,2.48105209E-2,
    X 3.24377334E-3,2.28338636E-4,7.8.255548E-6,1.08606937E-7,
    X 4.39934099E-10,2.22939385E-13/
  IF(N0.EQ.20) GO TO 120
  11: DO 22 I=1,5
    T(I)=-T10(6-I)
    22 WT(I)=WT10(6-I)
    DO 23 I=6,10
    T(I)=T10(I-5)
    23 WT(I)=WT10(I-5)
    GO TO 150
  12: DO 18 I=1,10
    T(I)=-T20(11-I)
    18 WT(I)=WT20(11-I)
    DO 19 I=11,20
    T(I)=T20(I-10)
    19 WT(I)=WT20(I-10)
  150 CONTINUE
  RETURN
  END

```

REGISTER ALLOCATION

1 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 15
 1 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 22

SUBROUTINE HERMIT

CDC 6600 FTN V3.0-P213 OPT=2 03/19/72 00.54.44.

PAGE

2

SYMBOLIC REFERENCE MAP

ENTRY POINTS	DEF LINE	REFERENCES
2 HERMIT	1	29

VARIABLES	SN	TYPE	RELOCATION
62 I		INTEGER	
0 NO		INTEGER	
0 I		REAL	F.P.
107 I10		REAL	F.P.
63 I20		REAL	
0 WT		REAL	
114 WT10		REAL	F.P.
75 WT20		REAL	

STATEMENT LABELS

DEF LINE	REFERENCES
24	22
27	25
17	15
20	18
15	14
22	21
28	

LOOPS LABEL	INDEX	FROM-TO	LENGTH	PROPERTIES
23 22	* I	15 17	48	INSTACK
34 23	I	18 20	38	INSTACK
43 18	* I	22 24	48	INSTACK
54 19	I	25 27	38	INSTACK

STATISTICS

PROGRAM LENGTH 1378 95

PAGE 1

COC 6600 FTN V3.0-P213 OP1=2 03/19/72 00.54.44.

SUBROUTINE COEF

```

SUBROUTINE COEF(N1,HG)
  DIMENSION L( , 9),HG(9,9)
  L(1,1) = 1
  L(2,2)=2
  L(2,1)=0
  L(3,3)=4
  L(3,2)=0
  L(3,1)=-2
  NOEG = N1 - 1
  DO 50 N=3,NOEG
    NN=N+1
    L(NN,NN)=2*L(NN-1,NN-1)
    L(NN,NN-1)=0
    L(NN,1)=-2*(N-1)*L(NN-2,1)
    NM=N-1
    DO 60 K=2,NM
      L(NN,NN-K)=2*L(NN-1,NN-1-K) - 2*(N-1)*L(NN-2,NN-K)
50  CONTINUE
    DO 60 K=1,N1
      DO 60 N=1,N1
        HG(K,N)=L(K,N)
60  CONTINUE
    WRITE(6,64)
    DO 63 I=1,N1
      WRITE(6,65)(HG(I,J),J=1,I)
63  CONTINUE
  64  FORMAT(*G,*,* HERMITE COEFFICIENTS*)
  65  FORMAT(*0*,9F10.0)
  RETURN
  END

```

SUMMARY OF CHANGES MADE BY THE OPTIMIZER
 52 WORDS OF INVARIANT RLJST REMOVED FROM THE LOOP STARTING AT LINE 16

SUBROUTINE COEF

SYMBOLIC REFERENCE MAP

ENTRY POINTS DEF LINE REFERENCES
2 COEF 1 29

VARIABLES SN TYPE REAL RELLOCATION
145 I INTEGER F.P.
146 J INTEGER
147 K INTEGER
147 L INTEGER ARRAY

REFS
2*25
25
2*17
2
3
14
11
10
16
4*12
9

25
DEFINED
2*21
12
4
17
14
20
DEFINED
2*13
19

DEFINED
24
25
DEFINED
14
5

1
21
16
2*17
6
17
2*21

19
21
7
12

8

11
1

141 N INTEGER
140 MDEG INTEGER
143 NM INTEGER
142 NM INTEGER
0 M1 INTEGER F.P.

FILE NAMES MODE
TAPE6 FMT

WRITES 23

STATEMENT LABELS DEF LINE REFERENCES

2 50 18
0 50 22
0 53 26
131 54 FMT 27
135 65 FMT 28

LOOPS LABEL INDEX FROM-TO ENGLISH PROPERTIES

24 50 * H 10 16 446 NOT INNER
51 58 * K 16 18 148 OPT
71 60 * K 19 22 118 NOT INNER
75 50 * N 20 22 28 INSTACK
106 63 * I 24 26 158 EXT REFS
111 * J 25 78 78 EXT REFS

STATISTICS
PROGRAM LENGTH 2778 191

SUBROUTINE CH3TRL
 CDC 6600 FIN V3.0-P213 OPT=2 03/19/78 09.54.44. PAGE 1

SUBROUTINE CH3TRL(K,A)
 DIMENSION A(17,17)

KN=K-1
 DO 20 N=1,K
 A(N,1)=1.0
 A(N,N)=1.0

2J CONTINUE

DO 21 N=2,KM

DO 21 M=2,N
 A(N+1,M) = A(N,M) + A(M,N-1)

21 CONTINUE

WRITE (6,51)

DO 23 I=1,K

WRITE(6,50) (A(I,J), J=1,I)

23 CONTINUE

50 FORMAT(' ',12F9.0)

51 FORMAT('1',5X,'BINOMIAL COEFFICIENTS')
 RETURN
 END

REGISTER ALLOCATION

1 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 4

SUBROUTINE CMBTRL

SYMBOLIC REFERENCE MAP

ENTRY POINTS DEF LINE REFERENCES
2 CMBTRL 1 18

VARIABLES	SN	TYPE	REAL	ARRAY	RELLOCATION F.P.	REFS
0 A		REAL				10
76 I		INTEGER				REFS
77 J		INTEGER				REFS
0 K		INTEGER				REFS
73 KN		INTEGER				REFS
75 M		INTEGER				REFS
74 N		INTEGER				REFS

FILE NAMES MODE
TAPES FMT

WRITES 12

14

STATEMENT LABELS DEF LINE REFERENCES

0 20	7	4
0 21	11	8
0 23	15	13
62 50	16	14
65 51	17	12

LOOPS	LABEL	INDEX	FROM-TO	LENGTH	PROPERTIES
22 20	N		4 7	28	INSTACK
25 21	* N		8 11	118	NOT INNER
31 21	H		9 11	28	INSTACK
42 23	* I		13 15	168	EXT REFS
45	* J		14	78	EXT REFS

STATISTICS
PROGRAM LENGTH 1128 74

PAGE 1

CDC 6600 FTN V3.0-P213 OPT=2 03/19/72 00.54.44.

SUBROUTINE GAM22

```

SUBROUTINE GAM22(N,A~MA)
  DIMENSION
  DC 153 I=2,N
  AI(1)=1.0
  DO 152 K=2,I
    AK=K
    AI(K)=AI(K-1)*AK
  152 CONTINUE
  FCIRL(1)=AI(1)
  153 CONTINUE
  FCIRL(1)=1.0
  DO 81 I=2,N
    DO 81 J=2,N
      AAMA(I,J)=2.0*(I-1)*2.0*(J-1)*FCIRL(I-1)*FCIRL(J-1)
      AAMA(I,J)=1.0/AAMA(I,J)
  81 CONTINUE
  AAMA(1,1)=1.0
  DO 82 K=2,N
    AAMA(K,1)=2.0*(K-1)*FCIRL(K-1)
    AAMA(K,1)=1.0/AAMA(K,1)
    AAMA(1,K)=AAMA(K,1)
  82 CONTINUE
  50 WRITE(6,50)
  50 FORMAT('3',5X,'NORMALIZING COEFFICIENTS*')
  51 WRITE(6,51)((AAMA(I,J),I=1,N),J=1,N)
  51 FORMAT('0',1P9E16.6)
  RETURN
END

```

SUMMARY OF CHANGES MADE BY THE OPTIMIZER
 13 WORDS OF INVARIANT RLST REMOVED FROM THE LOOP STARTING AT LINE 13

REGISTER ALLOCATION
 2 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 5

SYMBOLIC REFERENCE MAP

ENTRY POINTS DEF LINE REFERENCES
2 GAM22 1 27

VARIABLES SN TYPE REAL RELOCATION
 0 AAAA ARRAY F.P.

135 AI REAL REAL 20 25 21

133 AK REAL REAL 15 19 7

146 FCtrl REAL REAL 14 4 11

131 I INTEGER 19 9 25

134 J INTEGER 3*14 13 2*21

132 K INTEGER 3 18 25

0 N INTEGER 3*14 13 2*25

FILE NAMES MODE FMT WRITES 23 25

STATEMENT LABELS DEF LINE REFERENCES

117 50 FMT 24 23

124 51 FMT 26 25

0 81 16 12 13

0 82 22 18

0 152 8 5

0 153 10 3

LOOPS LABEL INDEX FROM-TO LENGTH PROPERTIES NOT INNER

15 153 * I 3 10 168 INSTACK EXT REFS NOT INNER

23 152 * K 5 8 38 EXT REFS EXT REFS EXT REFS

36 81 * I 12 16 258 EXT REFS EXT REFS EXT REFS

41 81 * J 13 16 238 EXT REFS EXT REFS EXT REFS

66 82 * K 18 22 158 EXT REFS EXT REFS EXT REFS

104 * J 25 68

STATISTICS PROGRAM LENGTH 1768 126

PAGE 1

CDC 6600 FTM V3.0-P213 OPT=2 03/19/72 00.54.44.

SUBROUTINE EIGEN

```

SUBROUTINE EIGEN(B,FKS)
  DIMENSION B(3),FKS(4),A(2,2),D(2,2),T(2,2)
  PI=3.14159266
  P2=PI*0.5

```

```

  IF(ABS(B(2))-T(1,0E-7) GO TO 15
  IF(ABS(B(1)-B(3))-LT(1,0E-7) GO TO 16
  PHI=0.5*ATAN(2.0*B(2)/(B(1)-B(3)))
  GO TO 20

```

```

15 PHI=0.0

```

```

10 GO TO 20

```

```

16 PHI=0.5*SIGN(-2,B(2))

```

```

20 CONTINUE

```

```

  A(1,1)=B(1)

```

```

  A(2,1)=B(2)

```

```

  A(1,2)=B(2)

```

```

  A(2,2)=B(3)

```

```

  T(1,1)=COS(PHI)

```

```

  T(2,1)=SIN(PHI)

```

```

  T(1,2)=T(1,1)

```

```

  T(2,2)=T(1,1)

```

```

  FKS(1)=T(1,1)

```

```

  FKS(2)=T(2,1)

```

```

  FKS(3)=T(1,2)

```

```

  FKS(4)=T(2,2)

```

```

  DO 25 I=1,2

```

```

    DO 25 J=1,2

```

```

      D(I,J)=0.0

```

```

      DO 25 K=1,2

```

```

        D(I,J)=A(I,K)*T(K,J) + D(I,J)

```

```

      25 CONTINUE

```

```

  25 CONTINUE

```

```

  T(1,2)=T(2,1)

```

```

  T(2,1)=D(1,1)

```

```

  DO 35 I=1,2

```

```

    DO 35 J=1,2

```

```

      A(I,J)=0.0

```

```

      DO 35 K=1,2

```

```

        A(I,J)=T(I,K)*D(K,J) + A(I,J)

```

```

      35 CONTINUE

```

```

  B(1)=A(1,1)

```

```

  B(3)=A(2,2)

```

```

  RETURN

```

```

  END

```

5

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CUC 6600 FIN V3.0-P213 OPT=2 03/19/72 00.54.44. PAGE 2

SUBROUTINE EIGEN

SYMBOLIC REFERENCE MAP

ENTRY POINTS	DEF LINE	REFERENCES
2 EIGEN	1	42

VARIABLES	SN	TYPE	RELLOCATION	REFS
144 A		REAL	ARRAY	DEFINED
0 B		REAL	ARRAY	REFS
150 0		REAL	ARRAY	REFS
143 01		REAL	ARRAY	REFS
0 FKS		REAL	ARRAY	REFS
140 I		INTEGER	ARRAY	REFS
141 J		INTEGER	ARRAY	REFS
142 K		INTEGER	ARRAY	REFS
137 PHI		REAL	ARRAY	REFS
135 PI		REAL	ARRAY	REFS
136 P2		REAL	ARRAY	REFS
154 T		REAL	ARRAY	REFS

EXTERNALS

TYPE	ARGS	REFERENCES
ATAN	1 LIBRARY	7
COS	1 LIBRARY	17
SIN	1 LIBRARY	18

INLINE FUNCTIONS

TYPE	ARGS	DEF LINE	REFERENCES
ABS	1 INTRIN	5	6
SIGN	2 INTRIN	11	

STATEMENT LABELS

DEF LINE	REFERENCES
33 15	9
35 16	11
41 20	12
0 25	30
0 35	39

LOOPS	LABEL	INDEX	FROM-TO	LENGTH	PROPERTIES
60 25	* I	25 33	2.8	NOT INNER	
61 25	* J	26 30	158	NOT INNER	
70 25	* K	28 30	38	INSTACK	
104 35	* I	34 39	2.8	NOT INNER	
105 35	* J	35 39	158	NOT INNER	
114 35	* K	37 39	38	INSTACK	

STATISTICS

PROGRAM LENGTH	2048
132	

SUBROUTINE CYCLE CDC 6600 FIN V3.0-P213 OPT=2 03/19/72 00.54.44. PAGE 1

```

SUBROUTINE CYCLE
  DIMENSION
    1 Y1(20,20),Y2(20,20),XKS(3),EKS(4),
    2 TAURK(20,20),TAUKL(20,20),HRK(8),MKL(16),
    3 BRL(9,9),Y1P(20),Y2P(20),
    4 CENMOM( 9,9 ),ALPHA1( 9),ALPHA2( 9),
    5 DENS(20,20),MT2(20,20),HH(20,20),BIN2(9,9),CENH(17,17),
  DIMENSION TT1PH(9),TT2PH(9),
  DIMENSION AAL1(9),AA21(9),AA12(9),AA22(9),
  REAL K1,K2,M1,M2,MON(3,3)
  REAL M1BAR,M2BAR
  DIMENSION XDAT(230,3)
  COMMON/NAME6/TT(20),MT(20),SAMA(9,3),HM(9,9),BINOM(17,17)
  COMMON/NAME8/A11,A22,Q22,R11,DELT,JSEED
  COMMON/NAME9/XONE,XTHO
  WRITE (6,652)
  WRITE (6,653) A11,A22,Q22,R11,DELT,NUM1,NO1
  PI=3.1415926536
  P12=2.0*PI
  NOH=NO1-1
  NSAMP=0
  SUMP1=0.0
  2 CONTINUE
    KOUNT = 1
    DELSQ = DELT**2
    CENMOM(1,1)=1.0
    CENMOM(1,2)=0.0
    CENMOM(2,1)=0.0
    CENMOM(2,2)=5.0
    CENH(1,1)=1.0
    CENH(2,1)=0.0
    CENH(1,2)=0.0
    AA11(1)=1.0
    AA21(1)=1.0
    AA12(1)=1.0
    AA22(1)=1.0
    BTA(1) = 1.0
    AR(1) = 1.0
    BL(1) = 1.0
    ALPHA1(1)=1.0
    ALPHA2(1)=1.0
    TT1PH(1) =1.0
    TT2PH(1) =1.0
    DO 30 I=1,NO1
      DO 30 J=1,NO1
        3 J BRL(I,J)=0.0
        00 35 I=1,NUM1
        00 35 J=1,NUM1
        35 MT2(I,J)=MT(I)*MT(J)
        DEV1=SQRT(A11)
        DEV2=SQRT(A22)
        DEV3=SQRT(R11)
        Y1CST=XONE

```

SUBROUTINE CYCLE

```

Y2EST=XTHO
CALL GAUSS(JSEED,DEV1,Y1EST,X1)
CALL GAUSS(JSEED,DEV2,Y2EST,X2)
ADAT(KOUNT,1)=X1
DEV2=SQR(T(Q22))
COSX1=COS(X1)
SINX1=SIN(X1)
CALL GAUSS(JSEED,DEV3,COSX1,Z1)
CALL GAUSS(JSEED,DEV3,SINX1,Z2)
***** INITIALIZE PARAMETERS *****
05 41 I=1,N01
00 41 J=1,N01
41 BRL(I,J)=9.0
BRL(1,1)=1.0
***** COMPUTE INITIAL MEAN AND VARIANCE BY LINEARIZING *****
K1=1.0/A11 + 1.0/R11
K1=1.0/K1
K2=A22
XKS(1)=K1
XKS(2)=0.
XKS(3)=K2
CALL EIGEN(XKS,XKS)
DEVK1=SQR(2.0*XKS(1))
DEVK2=SQR(2.0*XKS(3))
00 51 I=1,NUM1
Y1P(I) = DEVK1*Y1(I)
Y2P(I) = DEVK2*Y1(I)
51 CONTINUE
COSY1=COS(Y1EST)
SINY1=SIN(Y1EST)
BZ1=(Z1*SINY1-Z2*COSY1-Y1EST)/R11
M1=-BZ1*K1 + K1/A11*Y1EST
M2=Y2EST
00 56 I=1,NUM1
00 56 J=1,NUM1
MH(I,J)=1.0
Y1(I,J)=EKS(1)*Y1P(I) + EKS(3)*Y2P(J) + M1
Y2(I,J)=EKS(2)*Y1P(I) + EKS(4)*Y2P(J) + M2
56 CONTINUE
60 TO 425
***** MAIN RECYCLE POINT *****
70 CONTINUE
T11=-1.0/(EKS(1)*DELT - EKS(2))
T1SQ=T11**2
T12=EKS(1)*T11
T2SQ=T12**2
T13=1.0/(EKS(4)-EKS(3)*DELT)
T3SQ=T13**2
T4=EKS(3)*T13
AA=Q22 + T1SQ*S11
BB=T1SQ*S11/AA
BETA=-Q22/(S11*T1SQ)
CSQ= T13SQ*S22 + S11*T11SQ*(1.0 - BB)

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SUBROUTINE CYCLE

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115      CC = SQRT(CSQ)
          BSQCSQ=BB**2/CSQ
          V4 = TT1*V1BAR
          V5 = -BETA*TT1*V1BAR - TT3/BB*V2BAR
          V6=TT4/BB + BETA*TT2
          AR(2)= SQRT(BB)
          AR(3)=BB
          C3=1.0/(S22*Q22*TT3SQ + TT1SQ*TT3SQ*S11*S22 + S11*Q22*TT1SQ)
          BL(3)=S22*TT3SQ*C3*AA
          BL(2)= SQRT(BL(3))
          IF(TT1.LT.0.3)AR(2)=-AR(2)
          IF(TT3.LT.0.3)BL(2)=-BL(2)
          BTA(3) = Q22**2*C3
          BTA(2)=-SQRT(BTA(3))
          DO 120 K=4,N01
            BTA(K) = BTA(K - 1)*BTA(2)
            AR(K) = AR(K - 1)*AR(2)
            BL(K) = BL(K - 1)*BL(2)
          120 CONTINUE
          A2RT = 1.0/SQRT(2.0*AA)
          BCRT = BB/(SQRT(2.0)*CC)

130      X1=X1 + X2*DELT
          XDAT(KOUNT,1)=X1
          CALL GAUSS(JSEED,DEV2,X2,X2)
          COSX1=COS(X1)
          SINX1=SIN(X1)
          CALL GAUSS(JSEED,DEV3,COSX1,Z1)
          CALL GAUSS(JSEED,DEV3,SINX1,Z2)
          Y1EST=ETA1
          COSY1=COS(Y1EST)
          SINY1=SIN(Y1EST)
          BZ1=(Z1*SINY1-Z2*COSY1-Y1EST)/R11
          C11 = +TT2/AA + BSQCSQ*V6
          C22 = 1.0/AA + BSQCSQ
          C33 = 1.0/R11 + TT2SQ/AA + BSQCSQ*V6**2 - V4*TT2/AA + BSQCSQ*V5*V6
          C44= BZ1
          C55 = -V4/AA + BSQCSQ*V5
          K1=1.0/C33
          K2 = 1.0/C22
          RTK1 = SQRT(K1)
          RTK2 = SQRT(K2)
          RHO = -C11*RTK1*RTK2
          RHOIN = 1.0/(1.0 - RHO**2)
          XKS(1)=K1*RHOIN
          XKS(3)=K2*RHOIN
          XKS(2) = SQRT(XKS(1)*XKS(3))*RHO
          M1=-RHOIN*(RTK1*RTK2*RHO**355 + K1*C44)
          M2 =-RHOIN*(RTK1*RTK2*RHO*C44 + K2*C55)
          CALL EIGEN(XKS,EKS)
          ***** GENERATE NEW GRIE *****
          DEVK1=SQRT(2.0*XKS(1))
          DEVK2=SQRT(2.0*XKS(3))
          200 00 211 I=1,NUM1
155
160
165

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SUBROUTINE CYCLE

```

170      Y1P(I) = DEVK1*T(I)
      Y2P(I) = DEVK2*T(I)
      211 CONTINUE
      DO 212 I=1,NUM1
      DO 212 J=1,NUM1
      Y1(I,J)=EKS(1)*Y1P(I) + EKS(3)*Y2P(J) + M1
      Y2(I,J) = EKS(2)*Y1P(I) + EKS(4)*Y2P(J) + M2
      212 CONTINUE
      DO 215 I=1,NUM1
      DO 215 J=1,NUM1
      TAUUK(I,J) = -(TT2*Y1(I,J) - TT1*V1BAR)
      TAUKL(I,J) = 3CRT*(Y2(I,J) - BETA*TAURK(I,J) + (TT4*Y1(I,J) -
      1 TT3*V2BAR)/BB)
      TAUUK(I,J) = A2RT*(Y2(I,J) - TAUUK(I,J))
      215 CONTINUE
      DO 301 I=1,NUM1
      DO 301 J=1,NUM1
      HHH(I,J) = 0.0
      301 CONTINUE
      185      ***COMPUTATION OF POLYNOMIAL FUNCTION *****
      DO 251 IR=1,N01
      DO 251 K=1,IR
      251      BIN2(IR,K)=BIA(K)*BINOM(IR,K)
      DO 255 L=1,N01
      ITT=NO1+1-L
      DO 255 IR=1,ITT
      255      BRL(IR,L)=BRL(IR,L)*AR(IR)*BL(L)
      DO 311 I=1,NJ+1
      DO 311 J=1,NUM1
      HRK(I) = 2.0*TAURK(I,J)
      HKL(I) = 2.0*TAUKL(I,J)
      HRK(I2) = 4.0*TAURK(I,J)**2 - 2.0
      HKL(I2) = 4.0*TAUKL(I,J)**2 - 2.0
      DO 226 K=3,NOM
      XK = K - 1
      HRK(K) = 2.0*TAURK(I,J)*HRK(K-1) - 2.0*XK*HRK(K-2)
      HKL(K) = 2.0*TAUKL(I,J)*HKL(K-1) - 2.0*XK*HKL(K-2)
      226 CONTINUE
      DO 310 L=1,N01
      ITT=NO1+1-L
      DO 310 IR=1,ITT
      IF (BRL(IR,L).EQ.0.0) GO TO 310
      HHK=0.0
      DO 308 K=1,IR
      IN1= IR - K
      IN2 = K + L - 2
      IF (IN1.EQ.0.AND.IN2.EQ.0) GO TO 304
      IF (IN1.EQ.0) GO TO 305
      IF (IN2.EQ.0) GO TO 306
      HH1 = HRK(IN1) * HKL(IN2)
      HH1=HH1*BIN2(IR,K)
      GO TO 307
      304      HH1=1.0

```

```

SUBROUTINE CYCLE
      GO TO 307
235 305 HH1=HKL(IN2)      *BIN2(IR,K)
      GO TO 307
236 306 HH1=HRK(IN1)      *BIN2(IR,K)
      307 CONTINUE
      HHK = HH1 + d4K
237 308 CONTINUE
      HH2 = HHK*BRL(IR,L)
      HHH(I,J) = HH1(I,J) + HH2
238 310 CONTINUE
      IF(HHH(I,J).LT.0.0) HHH(I,J)=0..
239 311 CONTINUE
      425 CONTINUE
      C *** COMPUTATION OF EXPONENTIAL FUNCTION *****
240 312 CONTINUE
      SUMDEN = 0.0
      Z1R=Z1/R11
      Z2R=Z2/R11
      ZSQ=.5/R11
      Y1ES2=Y1EST**2
      Z1RS=Z1R*(COSY1+Y1EST*SINY1)+Z2R*(SINY1-Y1EST*COSY1)
      DO 411 I=1,NUM1
      DO 411 J=1,NUM1
      EXPON=EXP(Z1R*COS(Y1(I,J)) + Z2R*SIN(Y1(I,J))
      1 *Z1*Y1(I,J) +ZSQ*(Y1(I,J)**2+Y1ES2) - Z1RS )
      DENS(I,J) = EXPON*HHH(I,J)
      SUMDEN = DENS(I,J)*WT2(I,J) + SUMDEN
      411 CONTINUE
      SUMDI=1./SUMDEN
      DO 421 I=1,NUM1
      DO 421 J=1,NUM1
      DENS(I,J) = DENS(I,J)*SUMDI
      421 CONTINUE
      DO 426 I=1,NUM1
      DO 426 J=1,NUM1
      DENS(I,J) = DENS(I,J)*WT2(I,J)
      426 CONTINUE
      C ***** COMPUTE MEANS *****
      C ***** RAISE GRID POINTS TO POWERS *****
      DO 477 J=1,NUM1
      TAURK(1,J)=1.0
      TAUKL(1,J)=1.0
      TAURK(2,J)=Y1P(J)
      TAUKL(2,J)=Y2P(J)
      477 CONTINUE
      C ***** COMPUTE MOMENTS *****
      NK1=NO1
      DO 478 K=3,NK1
      DO 478 J=1,NUM1
      TAURK(K,J)=TAURK(K-1,J)*TAURK(2,J)
      TAUKL(K,J)=TAUKL(K-1,J)*TAUKL(2,J)
      478 CONTINUE
      DO 491 HH=1,NK1
      ITT=NK1+1-HH
245 250
255 260
265 270
275

```

SUBROUTINE CYCLE

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280      DO 491 NN=1,ITF
          GENHOM(MH,NN)=G.J
          DO 491 I=1,NU*1
              DO 491 J=1,NU*1
                  GENHOM(MH,NN)=TAURK(MH,I)*TAUKL(MH,J)*DENS(I,J) +CENHOM(MH,NN)
      491 CONTINUE
          MOM(2,1)=CENHOM(2,1)
          MOM(1,2)=CENHOM(1,2)
          M1BAR=MOM(2,1)*EKS(1)+MOM(1,2)*EKS(3)
          M2BAR=MOM(2,1)*EKS(2)+MOM(1,2)*EKS(4)
          ETA1=M1+M1BAR
          ETA2=M2+M2BAR
          XDAT(KOUNT,2)=ETA1
          ***** MOMENTS FROM CENTRAL MOMENTS *****
          IT1PH(2) =-CENHOM(2,1)
          IT2PH(2) =-CENHOM(1,2)
          DO 503 K=3,NK1
              IT1PH(K) =-CENHOM(2,1)*IT1PH(K-1)
              IT2PH(K) =-CENHOM(1,2)*IT2PH(K-1)
              IF(ABS(IT1PH(K))*.LT.1.0E-4) IT1PH(K) =0.0
              IF(ABS(IT2PH(K))*.LT.1.0E-4) IT2PH(K) =0.0
      503 CONTINUE
          DO 461 M=1,NK1
              ITT=NK1+1-M
              DO 461 N=1,ITF
                  GENH(M,N)=0.0
              DO 461 I=1,M
                  DO 461 J=1,N
                      GENH(M,N)=BINOM(M,I)*BINOM(N,J)* IT1PH(M-I+1) *
                      1 IT2PH(N-J+1) *CENHOM(I,J) + CENH(M,N)
      461 CONTINUE
          ***** COMPUTE COVARIANCE MATRIX *****
          XKS(1)=CENH(3,1)
          XKS(2)=CENH(2,2)
          XKS(3)=CENH(1,3)
          CALL EIGEN(XKS,FKS)
          GKS(1)=EKS(1)*FKS(1) + EKS(3)*FKS(2)
          GKS(2)=EKS(2)*FKS(1) + EKS(4)*FKS(2)
          GKS(3)=EKS(1)*FKS(3) + EKS(3)*FKS(4)
          GKS(4)=EKS(2)*FKS(3) + EKS(4)*FKS(4)
          DO 473 I=1,4
              EKS(I)=GKS(I)
      473 EKS(1)=GKS(1)
          V1BAR=ETA1*EKS(1)+ETA2*EKS(2)
          V2BAR=ETA1*EKS(3)+ETA2*EKS(4)
          M1=ETA1
          M2=ETA2
          S11=XKS(1)
          S22=XKS(3)
          P11 = EKS(1)**2*S11 + EKS(3)**2*S22
          P22 = EKS(2)**2*S11 + EKS(4)**2*S22
          P12 = EKS(1)*EKS(2)*S11 + EKS(3)*EKS(4)*S22
          WRITE (6,622) KOUNT
          WRITE (6,621) Z1,X1,ETA1,P11,P12,S11,EKS(1),EKS(3)
          WRITE (6,617)
          WRITE (6,621) Z2,X2,ETA2,P12,P22,S22,EKS(2),EKS(4)

```

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OPT=2

COC 660C FIN V3.0-P213

COORDINAT ROTATION OF MOMENTS *****

SUBROUTINE CYCLE

C*****

```

AA11(2)=FKS(1)
AA21(2)=FKS(3)
AA12(2)=FKS(2)
AA22(2)=FKS(4)
DO 492 J=1,N01
  AA11(J)=AA11(J-1)*FKS(1)
  AA21(J)=AA21(J-1)*FKS(3)
  AA12(J)=AA12(J-1)*FKS(2)
  AA22(J)=AA22(J-1)*FKS(4)
  IF (ABS(AA11(J)).LT.1.0E-40) AA11(J)=0.0
  IF (ABS(AA21(J)).LT.1.0E-40) AA21(J)=0.0
  IF (ABS(AA12(J)).LT.1.0E-40) AA12(J)=0.0
  IF (ABS(AA22(J)).LT.1.0E-40) AA22(J)=0.0

```

```

492 CONTINUE
50. CONTINUE

```

```

345
  CENHOM(1,1)=1.0
  CENHOM(2,1)=0.0
  CENHOM(1,2)=0.0
  CENHOM(2,2)=0.0
  CENHOM(3,1)=XKS(1)
  CENHOM(1,3)=XKS(3)
  IF (N01.EQ.3) GO TO 496
  DO 495 MM=1,N01
    ISS=5-MM
    IF (ISS.LT.1) ISS=1
    ITT=N01+1-MM
    DO 495 NN=ISS,ITT
      CENHOM(MM,NN)=C.0
      J1=MM+NN
      DO 495 IR=1,MM
        DO 495 IS=1,NN
          J2=IR+IS
          CENHOM(MM,NN)=8*INOM(MM,IR)*8*INOM(NN,IS)*AA11(MM-IR+1)*AA12(IR)*

```

```

355
  IF (ISS.LT.1) ISS=1
  ITT=N01+1-MM
  DO 495 NN=ISS,ITT
    CENHOM(MM,NN)=C.0
    J1=MM+NN
    DO 495 IR=1,MM
      DO 495 IS=1,NN
        J2=IR+IS
        CENHOM(MM,NN)=8*INOM(MM,IR)*8*INOM(NN,IS)*AA11(MM-IR+1)*AA12(IR)*

```

```

365
  1 AA21(NN+1-IS)*AA22(IS)*CENH(J1-J2+1,J2-1) + CENHOM(MM,NN)

```

```

495 CONTINUE
496 CONTINUE
  ALPHA1(2)=1.0/SQRT(2.0*S11)
  ALPHA2(2)=1.0/SQRT(2.0*S22)
  DO 505 I=1,N01
    ALPHA1(I)=ALPHA1(I-1)*ALPHA1(2)
    ALPHA2(I)=ALPHA2(I-1)*ALPHA2(2)

```

```

505 CONTINUE
  ALP=ALPHA1(2)*ALPHA2(2)/3.141592E

```

```

375
C ***** COMPUTATION OF COEFFICIENTS FOR EXPANSION *****

```

```

380
  BRL(1,1)=ALP
  DO 511 IR=1,N01
    ISS=5-IR
    IF (ISS.LT.1) ISS=1
    ITT=N01+1-IR
    DO 511 L=ISS,ITT
      DX=0.0
      DO 510 I=1,IR

```

385

SUBROUTINE CYCLE CDC 6600 FTN V3.0-P213 OPT=2 03/15/72 00.54.44. PAGE 8

```

00 510 J=1,L
IF(HM(IR,I).EQ.0.0) GO TO 513
IF(HM(L,J).EQ.0.0) GO TO 510
DX=DX + HM(IR,I)*HM(L,J)*ALPHA1(I)*ALPHA2(J)*CENHOM(I,J)
510 CONTINUE
511 BRL(IR,L)=ALP*GAMA(IR,L)*DX
CONTINUE
IF(KOUNT.EQ.N02) GO TO 550
KOUNT = KOUNT + 1
GO TO 70
552 FORMAT(*1*5X,A11*,11X,A22*,11X,A22*,10X,*R11*,11X,*JELT*,8X,
1*NO. OF INTG. PTS.,2X,*NO. OF TERMS*)
553 FORMAT(*1P5E14.6,2114/)
517 FORMAT(*12X,A22*,9X,A22*,11X,*ETA2*,14X,*P12*,11X,*P22*)
521 FORMAT(*15X,1P3E14.6,4X,1P2E14.6,4X,1P2E14.6,4X,1P2E14.6)
522 FORMAT(*0*,YTIME =*,13,3X,*Z1* 9X,*X1*,11X,*ETA1*,14X,*P11*,11X,
1*P12*)
1 14X,*EIGENVALUES*,11X,*EIGENVECTORS*)
550 CONTINUE
SUMP=0.0
00 1561 I=31,N02
XD=ABS(XDAT(I,1)-XDAT(I,2))
1498 CONTINUE
IF(XD.GT.PI) GO TO 1499
GO TO 1503
1499 XD=XD-PI
GO TO 1498
1500 CONTINUE
SUMP=(XD)*2 + SUMP
1501 CONTINUE
X=N02-30
SUMP=SUMP/X
NNS=NSAMP + 1
WRITE(6,1511)NNS, SUMP
1511 FORMAT(*0*,*SAMPLE PATH*,14,* STATISTICAL VARIANCE =*,1PE13.6)
XAA=XNSAMP+1.0
XNSAMP=NSAMP
SUMP1=(SUMP + XNSAMP*SUMP1)/XAA
OBSUN=ALOG10(SUMP1)*10.0
WRITE(6,1521)NNS,SUMP1,OBSUM
1521 FORMAT(*,*CUMULATIVE AVERAGE FJR*,14,* SAMPLE PATHS. STATISTICA
1L VARIANCE =*,1PE13.6,4X,* = *,1PE13.6,*08*)
NSAMP=NSAMP+1
IF(NSAMP.EQ.200) GO TO 1750
GO TO 2
1750 CONTINUE
RETURN
END

```

SUMMARY OF CHANGES MADE BY THE OPTIMIZER

28 WORDS OF INVARIANT RLIST REMOVED FROM THE LOOP STARTING AT LINE 175

19 WORDS OF INVARIANT RLIST REMOVED FROM THE LOOP STARTING AT LINE 303

32 WORDS OF INVARIANT RLIST REMOVED FROM THE LOOP STARTING AT LINE 362

REGISTER ALLOCATION

2 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 80

2 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 90

SUBROUTINE CYCLE
 3 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 125
 2 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 155
 3 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 170
 3 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 175
 3 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 201
 1 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 252
 1 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 261
 1 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 313
 2 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 362
 2 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 370

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SUBROUTINE CYCLE

SYMBOLIC REFERENCE MAP

ENTRY POINTS 1 CYCLE	DEF LINE 1	REFERENCES 432	RELOCATION	
VARIABLES 2101 AA	SN REAL	TYPE REAL	RELOCATION	
11025 AA11	REAL	REAL	ARRAY	REFS 148 107 337
11047 AA12	REAL	REAL	ARRAY	REFS 341 339
11036 AA21	REAL	REAL	ARRAY	REFS 343 338
11060 AA22	REAL	REAL	ARRAY	REFS 342 340 344
2172 ALP	REAL	REAL	ARRAY	REFS 378 2
5707 ALPHA1	REAL	REAL	ARRAY	REFS 371 2
5720 ALPHA2	REAL	REAL	ARRAY	REFS 372 2
2214 AR	REAL	REAL	ARRAY	REFS 117 121
0 A11	REAL	REAL	NAME8	REFS 16 19
2120 A2R1	REAL	REAL	NAME8	REFS 179 16
1 A22	REAL	REAL	NAME8	REFS 110 112
2103 B8	REAL	REAL	NAME8	REFS 177 108
2121 BCR1	REAL	REAL	NAME8	REFS 114 115
2104 BETA	REAL	REAL	NAME8	REFS 14 190
312 BINOH	REAL	REAL	NAME8	REFS 222 224
10211 BIN2	REAL	REAL	NAME8	REFS 2 120
2225 BL	REAL	REAL	NAME8	REFS 41 119
5375 BRL	REAL	REAL	NAME8	REFS 2 194
2110 BSQCSQ	REAL	REAL	NAME8	REFS 69 144
2203 BTA	REAL	REAL	NAME8	REFS 112 124
2071 BZ1	REAL	REAL	NAME8	REFS 124 87
2107 C	REAL	REAL	NAME8	REFS 131 304
20332 CENM	REAL	REAL	NAME8	REFS 2 33
2500 CENMUM	REAL	REAL	NAME8	REFS 32 280
2061 COSX1	REAL	REAL	NAME8	REFS 304 364
2067 COSY1	REAL	REAL	NAME8	REFS 294 31
2105 CSQ	REAL	REAL	NAME8	REFS 352 359
2123 C11	REAL	REAL	NAME8	REFS 63 138
2124 C22	REAL	REAL	NAME8	REFS 86 143
2116 C3	REAL	REAL	NAME8	REFS 111 112
				REFS 153 144
				REFS 150 145
				REFS 119 123
				REFS 118
				REFS 110
				REFS 61
				REFS 84
				REFS 141
				REFS 350
				REFS 29
				REFS 251
				REFS 364
				REFS 86
				REFS 143
				REFS 39
				REFS 123
				REFS 48
				REFS 190
				REFS 116
				REFS 117
				REFS 131
				REFS 109
				REFS 2364
				REFS 194
				REFS 128
				REFS 228
				REFS 147
				REFS 148
				REFS 39
				REFS 123
				REFS 364
				REFS 304
				REFS 293
				REFS 351
				REFS 352
				REFS 29
				REFS 251
				REFS 364
				REFS 86
				REFS 143
				REFS 39
				REFS 123
				REFS 48
				REFS 190
				REFS 116
				REFS 117
				REFS 131
				REFS 109
				REFS 2364
				REFS 194
				REFS 128
				REFS 228
				REFS 147
				REFS 148
				REFS 39
				REFS 123

SUBROUTINE CYCLE				JOC 660L FIN V3.0-P213 OPT=2. 03/19/72 00.54.44. PAGE 11			
VARIABLES	SN	TYPE	RELOCATION	REFS	DEFINED	146	147
2125 C33	REAL			149	DEFINED	146	147
2126 C44	REAL			158	159	DEFINED	148
2127 C55	REAL			158	159	DEFINED	148
2202 DBSUM	REAL			425	DEFINED	424	
2047 DELSQ	* REAL			27			
4 DELT	REAL			16	19	27	100
5731 OENS	REAL		NAME8	2	248	253	257
	REAL		ARRAY	247	253	257	280
2065 DEVK1	REAL			81	166	257	
2066 DEVK2	REAL			82	167	DEFINED	163
2052 DEV1	REAL			57	DEFINED	52	164
2053 DEV2	REAL			58	135	DEFINED	53
2054 DEV3	REAL			63	64	138	54
2173 DX	REAL			389	391	DEFINED	384
3701 EKS	REAL		ARRAY	2	77	DEFINED	389
				106	2*171	2*92	102
				2*314	2*318	2*172	2*312
				2*330	317	2*319	4*326
2122 ETAL	REAL			140	288	318	328
2160 ETA2	REAL			286	319	321	287
2153 EXPON	REAL			247	DEFINED	245	230
11015 FKS	REAL		ARRAY	9	311	2*312	2*315
				334	332	337	340
50 GAMA	REAL		NAME	14	391		314
11021 JKS	REAL		ARRAY	9	317	DEFINED	315
7371 HHH	REAL		ARRAY	2	229	231	183
				231	228	DEFINED	222
2140 HHK	REAL			226	226	DEFINED	222
2143 HH1	REAL			228	228	DEFINED	222
2144 HH2	REAL			229	DEFINED	228	222
5355 HKL	REAL		ARRAY	2	2*204	217	198
171 HH	REAL		NAME	14	387	388	197
5345 HRK	REAL		ARRAY	2	2*203	217	199
2050 I	INTEGER			48	2*51	68	91
				2*166	2*167	2*171	2*92
				183	198	199	3*179
				4*245	2*247	2*248	2*229
				2*317	2*372	387	3*304
				DEFINED	49	66	2*280
				174	195	80	165
				316	385	243	278
2141 IN1	INTEGER			214	215	406	212
2142 IN2	INTEGER			214	216	217	213
2134 IR	INTEGER			189	2*190	217	212
				224	228	3*194	211
				389	2*391	363	382
				363	3*364	DEFINED	361
2170 IS	INTEGER			363	3*364	DEFINED	208
2166 ISS	INTEGER			356	358	381	355
				381	381	DEFINED	356

SUBROUTINE CYCLE			CDC 6600 FYN V3.0-P213 OPT=2			03/19/72			00.54.44.			PAGE			12		
VARIABLES	SN	TYPE	RELOCATION														
2136 IIT	J	INTEGER															
2051 J	J	INTEGER	REFS	193	208	276	300	358	383	382	2*171						
			DEFINED	192	207	275	299	357	382	382							
			REFS	48	2*171	58	91	2*92	2*93	2*93							
			2*172	2*176	4*177	3*179	183	197	198	198							
			200	203	204	2*229	2*231	4*245	2*247	2*247							
			2*253	3*257	262	263	2*264	2*265	3*271	3*271							
			2*280	3*304	2*337	2*338	2*339	2*340	2*341	2*341							
			2*343	2*344	388	3*389	DEFINED	47	50	50							
			90	170	175	182	192	244	252	252							
			REFS	261	270	303	336	386	252	252							
			2*172	16	57	58	63	64	135	135							
5 JSEED	JSEED	INTEGER	REFS	139	DEFINED	360	327	393	394	394							
2167 J1	J1	INTEGER	REFS	364	DEFINED	363	3*190	202	3*283	3*283							
2171 J2	J2	INTEGER	REFS	2*364	DEFINED	363	224	2*271	2*272	2*272							
2117 K	K	INTEGER	REFS	2*126	2*127	2*128	125	189	201	201							
			212	213	218	222	125	189	201	201							
			2*294	2*295	2*296	DEFINED	125	189	201	201							
			269	292	292	292	125	189	201	201							
2046 KOUNT	KOUNT	INTEGER	REFS	59	134	288	327	393	394	394							
2933 K1	K1	REAL	DEFINED	26	394	74	2*87	151	155	155							
			REFS	11	72	149	156	159	155	155							
2034 K2	K2	REAL	DEFINED	71	72	152	156	159	155	155							
			REFS	11	76	152	156	159	155	155							
2135 L	L	INTEGER	DEFINED	73	150	152	156	159	155	155							
			REFS	192	3*194	207	209	213	228	228							
2161 M	M	INTEGER	REFS	388	2*391	DEFINED	191	206	383	383							
2156 MM	MM	INTEGER	REFS	299	301	302	4*304	DEFINED	298	298							
			REFS	275	277	3*280	355	357	359	359							
11071 MOM	MOM	REAL	REFS	361	DEFINED	274	354	282	283	283							
2035 M1	M1	REAL	REFS	11	2*284	2*285	DEFINED	DEFINED	87	87							
			REFS	11	92	171	286	DEFINED	158	158							
2037 M18AR	M18AR	REAL	REFS	12	286	DEFINED	284	DEFINED	88	88							
2036 M2	M2	REAL	REFS	11	93	172	287	DEFINED	159	159							
2040 M28AR	M28AR	REAL	REFS	12	287	DEFINED	289	DEFINED									
2162 N	N	INTEGER	REFS	311	303	4*304	DEFINED	306	299	299							
2155 NK1	NK1	INTEGER	REFS	269	274	275	292	298	299	299							
			DEFINED	268	268	268	360	362	4*364	4*364							
2157 NN	NN	INTEGER	REFS	277	3*280	359	360	362	4*364	4*364							
			DEFINED	276	358	359	360	362	4*364	4*364							
2177 NNS	NNS	INTEGER	REFS	419	425	DEFINED	418	DEFINED									
2043 NOM	NOM	INTEGER	REFS	201	DEFINED	22	418	DEFINED									
1 NO1	NO1	INTEGER	REFS	15	19	22	46	47	66	66							
			REFS	125	191	22	206	207	268	268							
2 NO2	NO2	INTEGER	REFS	353	357	370	379	382	336	336							
2044 NSAMP	NSAMP	INTEGER	REFS	15	393	406	416	DEFINED	23	23							
0 NUM1	NUM1	INTEGER	REFS	418	421	428	429	80	89	89							
			REFS	15	19	49	50	181	182	182							
			REFS	169	170	174	175	255	256	256							
			REFS	196	243	251	252	255	256	256							
			REFS	270	278	279	280	281	282	282							
2041 PI	PI	REAL	REFS	21	409	DEFINED	21	DEFINED	21	21							
2042 P12	P12	REAL	REFS	411	DEFINED	21	21	DEFINED	21	21							

SUBROUTINE CYCLE				CNC 6600 F1N V3.0-P213 OPT=2 03/19/72 00.54.44.										PAGE		14	
VARIABLES		SN	TYPE	RELOCATION													
2200	ANSAMP		REAL			164	311	322	323	351	352						
0	XONE		REAL			DEFINED	74	75	76	155	156		157				
1	XTMO		REAL			309	310						308				
2057	X1		REAL		NAME9	REFS	422	423	DEFINED	421							
					NAME9	REFS	17	55									
						REFS	17	56									
						REFS	57	59	61	62	133	134	136				
2060	X2		REAL			137	328	DEFINED	133								
2236	Y1		REAL			REFS	58	133	2*135	330	DEFINED		171				
2055	Y1EST		REAL			REFS	2	176	177	4*245	87	92	141				
				ARRAY		REFS	57	84	85	55	140	142					
2151	Y1ES2		REAL			143	241	2*242	DEFINED	241							
5516	Y1P		REAL			REFS	245	DEFINED	241								
				ARRAY		REFS	2	92	93	171	172	264					
						DEFINED	81	166									
3056	Y2		REAL		ARR.Y	REFS	2	177	179	DEFINED	93	172					
2056	Y2EST		REAL			REFS	58	88	DEFINED	56							
5542	Y2P		REAL		ARRAY	REFS	2	92	93	171	172	265					
						DEFINED	82	167									
2150	ZSQ		REAL			REFS	245	DEFINED	240								
2063	Z1		REAL			REFS	63	86	138	143	238	328					
2146	Z1R		REAL			REFS	242	245	DEFINED	238							
2152	Z1RS		REAL			REFS	245	DEFINED	242								
2064	Z2		REAL			REFS	64	86	139	143	239	330					
2147	Z2R		REAL			REFS	242	245	DEFINED	239							
FILE NAMES				MODE	WRITES	18	19	327	328	329	330	419	425				
TAPES				FMT													
EXTERNALS				TYPE	APGS	REFERENCES	424										
ALOG10			REAL		1 LIBRARY	61	84	141	245								
COS			REAL		1 LIBRARY	77	160	136									
EIGEN					2		311	311									
EXP			REAL		1 LIBRARY	245											
GAUSS					4	57	58	64	135	138	139						
SIN			REAL		1 LIBRARY	62	85	142	245								
SORT			REAL		1 LIBRARY	52	53	60	78	79	111	116	120				
						124	130	151	152	157	163	164	368				
						369											
INLINE FUNCTIONS				TYPE	ARGS	DEF LINE	REFERENCES	296	341	342	343	344	407				
ABS			REAL		1	INTRIN	295										
STATEMENT LABELS				DEF LINE	DEF LINE	REFERENCES	430										
34	2				25	48	46	47									
0	30				51	68	80	89									
0	35				83	94	395	125									
0	41				129	165	168										
0	51																
0	56																
224	70																
0	120																
0	200			INACTIVE	165												
0	211				168												

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SUBROUTINE CYCLE

STATEMENT LABELS	DEF LINE	REFERENCES	PROPERTY	NOT INNER
0 212	173	169	170	
0 215	180	174	175	
0 226	205	201		
0 251	190	188	189	
0 255	194	191	193	
0 301	184	181	182	
670 304	224	214		
672 305	222	215		
674 306	224	216		
676 307	225	219	221	223
0 308	227	211		
713 310	234	206		209
0 311	232	195	208	
0 411	249	243	196	
0 421	254	251	244	
734 425	233	95	252	
0 426	258	255	256	
0 461	306	298	300	
0 473	317	316	302	303
0 477	266	261		
0 478	273	269		
0 491	281	274	270	
0 492	345	336	276	279
0 495	366	354	351	
1516 496	367	353	358	362
0 500	346			
0 503	297	292		
0 505	373	370		
1564 510	390	385	386	388
0 511	392	379	383	
1610 550	434	393		
1753 617	399	329		
1761 621	401	328	330	
1767 622	401	327		
1736 652	396	18		
1750 653	398	19		
1617 1498	438	412		
1621 1499	411	409		
1623 1504	413	410		
0 1501	415	406		
2001 1511	420	419		
2010 1521	426	425		
1670 1753	431	429		
LOOPS LABEL	INDEX	FROM-TO	LENGTH	PROPERTY
52 30	* I	46 48	1JB	INSTACK
56 30	J	47 48	2B	NOT INNER
63 35	* I	49 51	11B	INSTACK
67 35	J	50 51	2B	NOT INNER
124 41	* I	66 68	10B	INSTACK
130 41	J	67 68	2B	NOT INNER
161 51	* I	80 83	3B	INSTACK
202 56	* I	89 94	21B	NOT INNER
210 56	J	90 94	11B	OPT

SUBROUTINE CYCLE		CDC 6600 FTN V3.0-P213 OPT=2		03/19/72	03.54.44.	PAGE	17
COMMON BLOCKS	LENGTH	MEMBERS - BIAS NAME(LENGTH)					
		3 R11 (1)					
) XONE (1)					
NAME9	2			4 DELT (1)	5 JSEED (1)		
				1 ATMO (1)			
STATISTICS							
PROGRAM LENGTH	123648	5364					
COMMON LENGTH	7668	512					

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```

FUNCTION  RANF
      FUNCTION RANF(NS,MODE)
      DIMENSION NS(2), NC(2)
      COMMON /RN/ N1, N2, MP, T1, T2
      DATA  M1, M2/24734, 158551/
      5      MODE=0 TO CONTINUE, OTHERWISE RESTART WITH
      6      INTEGER NUMBER NS(1)*2+18+NS(2)
      IF (MODE) 10, 100, 10
      1, N1=NS(1)
      N2=NS(2)
      T1=2.**(-18)
      T2=2.**(-36)
      MP=2**18
      10, DO 200 I=1,2
      GO TO (113,120),I
      110 K=N2*N2
      GO TO 190
      120 K=N1*N2+M2*N1+KD
      130 KD=K/MP
      200 NC(I)=K-KD*MP
      N1=NC(2)
      N2=NC(1)
      XN1=N1
      XN2=N2
      RANF=XN1*T1+XN2*T2
      RETURN
      END
25

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CUC 6630 FIN V3.0-P213 OPT=2 03/19/72 0J.54.44.

SUBROUTINE GAUSS

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SUBROUTINE GAUSS(JS,SD,AM,X)
  DIMENSION NST(2)
  COMMON /RN/ N1, N2, MC, I1, I2
  COMMON /GN/ TWOPI, J, XR(2)
  IF (J) 10, 13, 20
5  19 J=2
   TWOPI=6.283185307
   NST(1)=244734
   NST(2)=158551
   XR(1)=RANF(NST,1)
   GO TO 35
20 20 GO TO (30,40), J
30 31 J=2
   XR(1)=RANF(NST,0)
   XR(2)=RANF(NST,0)
   X1=SQR(ABS(-2.*ALOG(XR(1))))
   XR(2)=TWOPI*XR(2)
   XR(1)=X1*SIN(XR(2))
   XR(2)=X1*COS(XR(2))
   X=XR(1)*SD+XM
   RETURN
40 40 J=1
   X=XR(2)*SD+XM
   RETURN
END
25

```

CDC 6600 FTN V3.0-P213 OPT=2 03/19/72 00.54.44. PAGE 2

SUBROUTINE GAUSS

SYMBOLIC REFERENCE MAP

ENTRY POINTS DEF LINE REFERENCES 24
2 GAUSS 1 21

VARIABLES SN TYPE RELOCATION

1 J	INTEGER	GN	REFS	4	5	12	DEFINED	6	13	22
0 JS	INTEGER	*UNUSED	DEFINED	1						
2 MC	INTEGER	RN	REFS	3						
100 NST	INTEGER	ARRAY	REFS	2	10	14	15	DEFINED	8	9
0 N1	INTEGER	RN	REFS	3						
1 N2	INTEGER	RN	REFS	3						
0 SD	REAL	F.P.	REFS	20	23	DEFINED	1			
0 THOPI	REAL	GN	REFS	4	17	DEFINED	7			
3 T1	REAL	RN	REFS	3						
4 T2	REAL	RN	REFS	3						
0 X	REAL	RN	REFS	3						
0 XM	REAL	F.P.	DEFINED	1	20	23				
2 XR	REAL	F.P.	REFS	20	23	DEFINED	1	19	20	23
	REAL	GN	REFS	4	16	17		18	19	
	REAL	ARRAY	DEFINED	10	14	15		18		
77 X1	REAL	REFS	REFS	18	19	DEFINED	16			

EXTERNALS

EXTERNALS	TYPE	ARGS	REFERENCES
ALOG	REAL	1	LIBRARY 16
ATAN	REAL	1	LIBRARY 7
COS	REAL	1	LIBRARY 19
RANF	REAL	2	LIBRARY 10
SIN	REAL	1	LIBRARY 18
SQRT	REAL	1	LIBRARY 16

INLINE FUNCTIONS

ABS	TYPE	ARGS	DEF LINE	REFERENCES
	REAL	1	INTRIN	16

STATEMENT LABELS

STATEMENT LABELS	INACTIVE	DEF LINE	REFERENCES
0 10	6	2*5	
17 20	12	5	
25 30	13	12	
31 35	15	11	
54 40	22	12	

COMMON BLOCKS

COMMON BLOCKS	LENGTH	MEMBERS - BIAS NAME(LENGTH)
RN	5	3 N1 (1)
GN	4	3 T1 (1)
		3 THOPI (1)

STATISTICS

PROGRAM LENGTH	1028	66
COMMON LENGTH	118	9

1 N2	(1)
4 T2	(1)
1 J	(1)
2 MC	(1)
2 XR	(2)

EQUILIBRIUM CONTINUOUS POSITION VARIANCE = 1.000000E-04
 FILTER TIME CONSTANT = 2.714418E-01

DELTA = 2.714418E-02

DISCRETE DRIVING VARIANCE = 2.714418E-04

DISCRETE NOISE VARIANCE = 5.000000E-04

VARIANCE OF INIT. POS. AND VEL. 1.000000E-04 2.714418E-03

CONTINUOUS DRIVING VARIANCE = 1.000000E-02

CONTINUOUS NOISE VARIANCE = 1.357209E-05

HERMITE COEFFICIENTS

1.			
0.	2.		
-2.	0.	4.	
0.	-12.	0.	8.
12.	0.	-48.	0.
			16.

BINOMIAL COEFFICIENTS

1.			
1.	1.		
1.	2.	1.	
1.	3.	3.	1.
1.	4.	6.	4.
			1.

NORMALIZING COEFFICIENTS

1.000000E+00	5.060000E-01	1.250000E-01	2.083333E-02	2.604167E-03	5.000000E-01	2.500000E-01	6.250000E-02	1.041667E-02
1.302083E-03	1.250000E-01	6.250000E-02	1.562500E-02	2.604167E-03	3.255208E-04	2.083333E-02	1.041667E-02	2.604167E-03
4.340278E-04	5.425347E-05	2.604167E-03	1.302083E-03	3.255208E-04	5.425347E-05	6.781684E-06		

A11				Q22				R11				DEL1				NO. OF INTG. PTS.				NO. OF TERMS			
1.00000E-04				2.714418E-03				2.714418E-04				2.714418E-02				10				5			
TIME = 1				Z1				E1A1				P11				EIGENVALUES				EIGENVECTORS			
9.865661E-01				1.350673E-02				5.187855E-03				8.351885E-05				8.351885E-05				1.300000E+00			
3.105892E-02				7.803262E-03				1.215371E-16				0.				2.714418E-03				1.000000E+00			
TIME = 2				Z1				E1A1				P11				EIGENVALUES				EIGENVECTORS			
1.017557E+00				1.381854E-02				7.643203E-03				7.283972E-05				6.25732E-05				9.997667E-01			
2.213402E-02				2.799281E-03				2.115449E-03				6.27732E-05				2.976448E-03				-2.159844E-02			
TIME = 3				Z1				E1A1				P11				EIGENVALUES				EIGENVECTORS			
9.79979E-01				1.389452E-02				8.731258E-03				6.798722E-05				1.244222E-04				9.992207E-01			
1.512597E-02				1.143564E-02				4.001549E-03				1.244222E-04				3.212883E-03				-3.94707E-02			
TIME = 4				Z1				E1A1				P11				EIGENVALUES				EIGENVECTORS			
9.601708E-01				1.420492E-02				1.27156E-02				6.715242E-05				1.843345E-04				9.984917E-01			
1.914637E-02				2.019664E-02				7.330894E-03				1.843345E-04				3.409401E-03				-5.49.331E-02			
TIME = 5				Z1				E1A1				P11				EIGENVALUES				EIGENVECTORS			
1.15889E+00				1.475314E-02				1.644970E-02				6.857135E-05				2.382763E-04				9.976832E-01			
5.413595E-02				1.598442E-02				2.865091E-02				2.382763E-04				3.546701E-03				-6.803053E-02			
TIME = 6				Z1				E1A1				P11				EIGENVALUES				EIGENVECTORS			
9.82545E-01				1.518702E-02				2.103715E-02				7.218435E-05				2.87.813E-04				9.968005E-01			
4.331337E-02				2.53935E-02				4.350244E-02				2.07.813E-04				3.622385E-03				-7.992921E-02			
TIME = 7				Z1				E1A1				P11				EIGENVALUES				EIGENVECTORS			
9.92736E-01				1.587647E-02				1.772359E-02				7.667975E-05				3.269138E-04				9.959103E-01			
-7.298512E-03				2.57272E-02				2.460627E-02				3.269138E-04				3.650610E-03				-9.34794E-02			
TIME = 8				Z1				E1A1				P11				EIGENVALUES				EIGENVECTORS			
1.31279E+00				1.657482E-02				1.849129E-02				8.09.804E-05				3.549052E-04				9.95.801E-01			
1.958554E-02				2.091373E-02				2.504391E-02				3.549052E-04				3.61.165E-03				-9.907418E-02			
TIME = 9				Z1				E1A1				P11				EIGENVALUES				EIGENVECTORS			
1.00722E+00				1.714251E-02				2.152361E-02				8.518057E-05				3.751475E-04				9.942862E-01			
3.371209E-02				2.306250E-02				3.584514E-02				3.751475E-04				3.539174E-03				-1.1067473E-01			
TIME = 10				Z1				E1A1				P11				EIGENVALUES				EIGENVECTORS			
1.013337E+00				1.776852E-02				2.134114E-02				8.871063E-05				3.855011E-04				9.935888E-01			
1.18557E-02				2.349414E-02				2.671838E-02				3.86.011E-04				3.441526E-03				-1.13.546E-01			
TIME = 11				Z1				E1A1				P11				EIGENVALUES				EIGENVECTORS			
9.687519E-01				1.040625E-02				1.976813E-02				9.218325E-05				3.941984E-04				9.929463E-01			
1.345221E-02				3.153893E-02				2.54795E-02				3.94.984E-04				3.346408E-03				-1.185649E-01			
TIME = 12				Z1				E1A1				P11				EIGENVALUES				EIGENVECTORS			
1.104325E+00				1.926235E-02				2.446346E-02				9.403966E-05				3.932947E-04				9.924799E-01			
4.240717E-02				2.498145E-02				3.784114E-02				3.93.947E-04				3.234420E-03				-1.224077E-01			

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Additional Appendix D.

A Point-Mass Program for Implementing the Interpolating Version of the Cyclic Phase Demodulator

This program simulates the cyclic interpolating demodulator,
described in Appendix C of Chapter VII.

The program was compiled on one of the Kirtland AFB CDC 6600's
using the FTN compiler. The code-optimization algorithm was employed.
The random number generator is identical to the one described in
Chapter IV.

Programmer: C. Hecht

PROGRAM MAIN PAGE 1

CDC 660 F1N V3.0-P213 OPT=2 03/22/72 16.47.23.

```

PROGRAM MAIN(INPUT,OUTPUT,TAPES=INPUT,TAPE6=OUTPUT)
COMMON OENS1(21,105),DENS2(21,105),KLOM(21,105),KUP(21,105),
1 PROP(21,105)
COMMON /RN/ YZZ(1), XNZ(2)
COMMON /GN/ JZZ(1), JGAUSS, XZZ(2)
DIMENSION XDAT(100,5),XHAT(2)
DIMENSION
1 Y1(81),Y2(287),SINY(81),COSY(81),EXON(30),EXOON(15)
DIMENSION XDATM(100,5)
DIMENSION Y3(15)
DIMENSION SEVS(81)
DIMENSION Q(2,2),PBAR(2,2),PN(2,2),ANK(2),F(2,2),POUHY(2,2),
1 POUHY2(2,2),EIGVEC(2,2),PNF(2,2)
LOGICAL LOW,JP
LOGICAL SPLSI, SPL
DATA NBL, NAS/1n, 1H*/
2 C UE ,J
3 4,15,651)Y1EST,Y2EST,ALP110,DELTA,Q22C,NUM1,NUM2,N02,JSEED
SPL=.FALSE.
IF (JSEED.GT.0) GO TO 6665
SPL=.TRUE.
JSEED=-JSEED
6666 CONTINUE
P110=10.**((ALP110/10.))
Q22=Q22C**(.25)
RX=(P110/(SQRT(2.0)*Q22))**((4.0/3.0)
FTC=SQRT(2.0)*RX**(.25)/Q22
DELTA=DELTA*FTC
Q22=Q22C*DELTA
R11=RX/DELTA
A11=11.**((ALP110*1.4)/10.))
A22=2.*P110/(FTC*FTC)
WRITE (6,652) P110,ALP110,FTC,DELTA,Q22,R11
WRITE (6,653) Q22C,RX
653 FORMAT('0','CONTINUOUS DRIVING VARIANCE =',IPE14.6/
1 * CONTINUOUS NOISE VARIANCE =',IPE14.6)
651 FORMAT(5E11.4,4I5)
652 FORMAT('1','EQUILIBRIUM CONTINUOUS POSITION VARIANCE =',IPE14.6
1 * (IPE14.6* DB)* /
1 * FILTER TIME CONSTANT =*
1 * DELTA =*
1 * DISCRETE DRIVING VARIANCE =*
1 * DISCRETE NOISE VARIANCE =*
XMCN=0.
XMCN=0.
XMCN=J.
XMCN=0.
XMCN=0.
XMCN=J.
NHC=1
DO 8999 I=1,N02
DO 8999 J=1,N
8999 XDATA(I,J)=0.

```

```

PROGRAM      MAIN
CJC 66J: FTN V3.0-2213 OPT=2   C3/22/72   16.47.23.   PAGE      2

900. KOUNT=1
   SPLST=.FALSE.
   IF (NMC.EQ.JSEED.OR.SPL) SPLST=.TRUE.
   DELSQ=DEL*PI
   PI=3.1415926536
   PI2=2.0*PI
   PI0LT=PI/DEL
   PINV=1./PI
   PI20LT=2.*PI0LT
   U1=NUM1
   U2=NUM2
   Q(1,1)=1./J
   Q(2,2)=Q22
   XHAT(1)=Y1EST
   XHAT(2)=Y2EST
   XHAT1=Y1EST
   XHAT2=Y2EST
   PN(1,1)=A11
   PN(2,2)=A22
   PN(1,2)=0.
   PN(2,1)=0.
   R=RI1
   DEN = 1./((PN(1,1) + K)
   F(1,1)=1.
   F(1,2)=DEL
   F(2,1)=J.
   F(2,2)=1.
   A=DEL*P*INV*SQRT(1.-Q*Q22)
   IA=A*.5
   IY2=U2/PI20LT*SQRT(5.-Q*Q22) + .5
   IF (NMC.LE.1) WRITE (6,931) I2,IY2
901 FORMAT(*,*,1X,*NUMBER OF CYCLES ON EACH SIDE =*,I4,6X,
   1 *NUMBER OF POINTS ON EACH SIDE =*,I4)
   ***** GR10 Y1 AND Y2 *****
   EDG1=PI/U1
   EDG2=PI0LT/U2
   DO 4J I=1,NUM1
   X=I-1
   X=X/U1
4: Y1(I)=-PI + X*PI2 +EDG1
   DO 50 I=1,NUM2
   X=I-1
   X=X/U2
5: Y2(I)=-PI0LT + X*PI20LT + EDG2
   DO 51 I=1,IY2
   X=I
   X=X/U2
51 Y3(I)=X*PI20LT
   IF (NMC.GT.1) GO TO 9689
   WRITE(6,913)
   WRITE(6,912)(Y1(I),I=1,NUM1)
   WRITE(6,914)
   WRITE(6,912)(Y2(I),I=1,NUM2)
9689 CONTINUE
912 FORMAT(*,*,I10E12.4)

```

PROGRAM NAME

```

913 FORMAT('0',*Y1,GRID*)
914 FORMAT('0',*Y2,GRID*)
00 55 I=1,NUM1
COSY(I)=COS(Y1(I))
55 SINY(I)=SIN(Y1(I))
***** SUM OF EXPONENTIALS *****
IB=2*IA+1
IF(IY2.EQ.0) GO TO 153
00 151 I=1,IY2
EXDON(I)=3.0
00 150 IR=1,IB
IC=-IA-1*IK
R=IC
R=PI2DLF*R
15. EXDON(I)=EXDON(I) + EXP(-J.5/422*(Y3(I)+R)**2)
00 152 I=1,IY2
EXPON(I)=EXDON(IY2+1-I)
EXPON(I,IY2+1)=EXDON(I)
152 CONTINUE
153 CONTINUE
EXPON(IY2+1)=1.0
IY2=2*IY2+1
WRITE(6,912)(EXPON(I),I=1,IY2)
***** INITIAL DENSITY *****
SUM=3.0
00 161 I=1,NJM1
00 162 J=1,NJM2
DENS1(I,J)=EXP(-J.5/A11*Y1(I)**2-.5/A22*Y2(J)**2)
16. SUM=SUM+DENS1(I,J)
00 165 I=1,NJM1
00 165 J=1,NJM2
DENS1(I,J)=SUM*DENS1(I,J)
***** INTEGER ARRANGEMENT *****
RATIO=U1/U2
00 251 I=1,NUM1
AI=I
00 252 J=1,NJM2
AJ=J
AK=AI-RATIO*AJ + .5*(RATIO+U1)
KL=AK
IF(AK.LT.0.0) KL=KL-1
KU=KL+1
REALK=KL
PROP(I,J)=AK-R*ALK
KLOW(I,J)=KL
IF(KL.LT.1) KLOW(I,J)=KL+NUM1
IF(KL.GT.NUM1) KLOW(I,J)=KL-NUM1
KUP(I,J)=KU
IF(KU.LT.1) KUP(I,J)=KU+NJM1
IF(KU.GT.NUM1) KUP(I,J)=KU-NJM1
25. CONTINUE
DEV1= SQRT(A11)
DEV2= SQRT(A22)
DEV3= SQRT(R11)

```

PROGRAM MAIN

```

170 CALL GAUSS(JSEED,DEV1,YTEST,X1 )
    XOAT(KOUNT,1)=X1
    CALL GAUSS(JSEED,DEV2,YTEST,X1 )
    CALL GAUSS(JSEED,DEV3,COS(X1 ),Z1)
    CALL GAUSS(JSEED,DEV3,SIN(X1 ),Z2)
    DEVQ2= SQRT(Q22)
    R=R11
    NUNNL=0
    NUNPL=0
    CS=.75*PI
    ENL=ABS(XOAT(1,1)-XOAT(1,2))
    EPL=ABS(XOAT(1,1)-XOAT(1,4))
    IF (ENL.GT.CS) NUNNL=NUNNL+1
    IF (EPL.GT.CS) NUNPL=NUNPL+1
    DO 5656 I=1,NJM1
    XOEN(I)=0.
    DO 5656 J=1,NJM2
    5656 XOEN(I)=XOEN(I)+OENS1(I,J)
    GO TO 470
    450 CONTINUE
    X1=X1 + X2*DELT
    XOAT(KOUNT,1)=X1
    CALL GAUSS(JSEED,DEVQ2,X2,X2)
    CALL GAUSS(JSEED,DEV3,COS(X1 ),Z1)
    CALL GAUSS(JSEED,DEV3,SIN(X1 ),Z2)
    5 *** RICCATI EQUATION UPDATE *****
    POUNY(1,1)=(R*(PN(1,1)+2.)*PN(1,2)*DELT)-PN(1,2)**2*OEL34)*XOEN
    1 + PN(2,2)*DELS4
    POUNY(1,2)=PN(1,2)*(R-PN(1,2)*DELT)*XOEN + PN(2,2)*DELT
    POUNY(2,2)=-PN(1,2)**2*XOEN + PN(2,2) + 1(2,2)
    PN(1,1)=POUNY(1,1)
    PN(1,2)=POUNY(1,2)
    PN(2,2)=POUNY(2,2)
    PN(2,1) = PN(1,2)
    XEN =1.0/(PN(1,1) + R)
    200 CONTINUE
    700 UPDATE DENSITY FUNCTION *****
    SUM=0.0
    DO 285 I=1,NJM1
    285 SENS(I)=EXP(-.5/R11*((Z1-COSY(I))**2+( Z2-SINY(I))**2) )
    DO 310 J=1,NJM2
    LOW=FALSE.
    UP=FALSE.
    K1=J-IY2
    K2=J+IY2
    IF(K1.LT.1)LOW=.TRUE.
    IF(K2.GT.NJM2)UP=.TRUE.
    IF(LOW) K1=K1+NUM2
    IF(UP) K2=K2-NJM2
    DO 311 I=1,NJM1
    MN=0
    OUM=0.0
    IF(LOW) GO TO 301
    IF(UP) GO TO 305
    DO 300 L=K1,K2

```

PROGRAM MAIN

```

225      MM=MM+1
        KL=KLOW(I,L)
        KU=KUP(I,L)
        DUM=EXPON(MM)*(DENS1(KL,L)*(1.-PROP(KL,L))+DENS1(KU,L)*PROP(KL,L))
        1 + DUM
300 CONTINUE
        GO TO 315
230      MM=MM+1
        KL=KLOW(I,L)
        KU=KUP(I,L)
        DUM=EXPON(MM)*(DENS1(KL,L)*(1.-PROP(KL,L))+DENS1(KU,L)*PROP(KL,L))
        1 + DUM
303 CONTINUE
        GO TO 315
235      MM=MM+1
        KL=KLOW(I,L)
        KU=KUP(I,L)
        DUM=EXPON(MM)*(DENS1(KL,L)*(1.-PROP(KL,L))+DENS1(KU,L)*PROP(KL,L))
        1 + DUM
304 CONTINUE
        GO TO 315
240      MM=MM+1
        KL=KLOW(I,L)
        KU=KUP(I,L)
        DUM=EXPON(MM)*(DENS1(KL,L)*(1.-PROP(KL,L))+DENS1(KU,L)*PROP(KL,L))
        1 + DUM
305 CONTINUE
        GO TO 315
245      MM=MM+1
        KL=KLOW(I,L)
        KU=KUP(I,L)
        DUM=EXPON(MM)*(DENS1(KL,L)*(1.-PROP(KL,L))+DENS1(KU,L)*PROP(KL,L))
        1 + DUM
306 CONTINUE
        GO TO 315
250      MM=MM+1
        KL=KLOW(I,L)
        KU=KUP(I,L)
        DUM=EXPON(MM)*(DENS1(KL,L)*(1.-PROP(KL,L))+DENS1(KU,L)*PROP(KL,L))
        1 + DUM
307 CONTINUE
        GO TO 315
255      MM=MM+1
        KL=KLOW(I,L)
        KU=KUP(I,L)
        DUM=EXPON(MM)*(DENS1(KL,L)*(1.-PROP(KL,L))+DENS1(KU,L)*PROP(KL,L))
        1 + DUM
308 CONTINUE
        GO TO 315
260      MM=MM+1
        KL=KLOW(I,L)
        KU=KUP(I,L)
        DUM=EXPON(MM)*(DENS1(KL,L)*(1.-PROP(KL,L))+DENS1(KU,L)*PROP(KL,L))
        1 + DUM
309 CONTINUE
        GO TO 315
265      MM=MM+1
        KL=KLOW(I,L)
        KU=KUP(I,L)
        DUM=EXPON(MM)*(DENS1(KL,L)*(1.-PROP(KL,L))+DENS1(KU,L)*PROP(KL,L))
        1 + DUM
310 CONTINUE
        GO TO 315
270      MM=MM+1
        KL=KLOW(I,L)
        KU=KUP(I,L)
        DUM=EXPON(MM)*(DENS1(KL,L)*(1.-PROP(KL,L))+DENS1(KU,L)*PROP(KL,L))
        1 + DUM
311 CONTINUE
        GO TO 315
275      MM=MM+1
        KL=KLOW(I,L)
        KU=KUP(I,L)
        DUM=EXPON(MM)*(DENS1(KL,L)*(1.-PROP(KL,L))+DENS1(KU,L)*PROP(KL,L))
        1 + DUM
312 CONTINUE
        GO TO 315

```

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PROGRAM MAIN

```

280      SHAT=SHAT+DEVS1(I,J)*SINY(I)
        CHAT=CHAT+DENS1(I,J)*COSY(I)
36.    CONTINUE
        CXHAT=ATAN(SHAT/CHAT)
        IF(CHAT.GT.0.0) GO TO 367
        IF(SHAT.GT.0.0) CXHAT=CXHAT + PI
        IF(SHAT.LT.0.0) CXHAT=CXHAT-PI
367    PLOSS=2.0*(1.0-SQRT(SHAT**2+CHAT**2))
        XDAT(KOUNT,2)=CXHAT
        XDAT(KOUNT,3)=PLOSS
        PNF(1,1)=PN(1,1)*2*DEN
        PNF(1,2)=PN(1,2)*2*DEN
        PNF(2,1)=PNF(1,2)
        PNF(2,2)=PNF(2,2) - PN(1,2)**2*DEN
        ***    FILTER UPDATE    ***
        SINF1=SIN(XHAT1)
        COSF1=COS(XHAT1)
        XHAT(1)=XHAT1+DEN*(-PN(1,1)*SINF1+Z1*PN(1,1)*COSF1+Z2)
        XHAT(2)=XHAT2+DEN*(-PN(1,2)*SINF1+Z1*PN(1,2)*COSF1+Z2)
        XIFF=XHAT(1)
84    CONTINUE
        IF(XIFF.GT.PI) GO TO 88
        IF(XIFF.LT.-PI) GO TO 89
        GO TO 93
88    XIFF=XIFF-PI2
        GO TO 84
89    XIFF=XIFF+PI2
        GO TO 84
93    CONTINUE
        ***    PREDICTOR UPDATE    **
        XHAT1=XHAT(1) + DELT*XHAT(2)
        XHAT2=XHAT(2)
        XDAT(KOUNT,4)=XHAT(1)
        XDAT(KOUNT,5)=PNF(1,1)
        ENL=ABS(XDAT(KOUNT,1)-XDAT(KOUNT,2))
        EPL=ABS(XDAT(KOUNT,1)-XDAT(KOUNT,4))
        IF (ENL.GT.CS) NUNNL=NUNNL+1
        IF (EPL.GT.CS) NUNPL=NUNPL+1
        DO 5676 I=1,NUNH1
        XDEN(I)=0.
5676    XDEN(I)=XDEN(I)+DENS1(I,J)
        XHM0=X1
184    CONTINUE
        IF(XHM0.GT.PI) GO TO 188
        IF(XHM0.LT.-PI) GO TO 189
        GO TO 190
188    XHM0=XHM0-PI2
        GO TO 184
189    XHM0=XHM0+PI2
        GO TO 184
190    CONTINUE
222    IF (KOUNT.EQ.M02) GO TO 505
        KOUNT=KOUNT + 1
        GO TO 450

```


PROGRAM MAIN

```

207 FORMAT(*1*,10E15.6)
505 CONTINUE
IF (SPLST) WRITE (6,90J00) NMC
30000 FORMAT(*1*35X*MONTE CARLO SAMPLE PATH AVERAGES FOR PATH NO. *15/
1 *0*2X*N*8X*N-E*11X*AVNLE*1JX*AVNLV*5X*NLCS*6X*PLLE*10X*AVPLLE*
2 9X*AVPLLV*8X*AVCPLLV*9X*PLCS*//)
CS=.75*PI
CSN=0.
GSP=0.
DO 6200 I=1,N02
CALL MOD2PI(XDAT(I,1)-XDAT(I,2),ERRN)
ICSNLF=NBL
IF (ABS(ERRN).LE.CS) GO TO 6000
ICSNLF=NAS
CSN=CSN+1./N02
530. XDAT(I,1)=AVG(XDAT(I,1),ERRN,NMC)
XDAT(I,2)=AVG(XDAT(I,2),ERRN,NMC)
CALL MOD2PI(XDAT(I,1)-XDAT(I,4),ERRP)
ICSPLL=NBL
IF (ABS(ERRP).LE.CS) GO TO 6100
ICSPLL=NAS
GSP=GSP+1./N02
610. XDAT(I,3)=AVG(XDAT(I,3),ERRP,NMC)
XDAT(I,4)=AVG(XDAT(I,4),ERRP,NMC)
XDAT(I,5)=AVG(XDAT(I,5),XDAT(I,5),NMC)
SHN=XDAT(I,1)**2
SMP=XDAT(I,3)**2
XOMSN=XDAT(I,2)-SHN
XOMSP=XDAT(I,4)-SMP
IF (SPLST) WRITE (6,90J00) I,ERRN,XDAT(I,1),XOMSN,
1 ICSNLF,ERRP,XDAT(I,3),XOMSP, XDAT(I,5),ICSPLL
90J00 FORMAT(*15,3E15.7,2X,A1,2X,4E15.7,2X,A1)
6200 CONTINUE
IF (SPLST) WRITE (6,90200) CS, CSN, GSP
90200 FORMAT(*0*29X*CYCLE SLIP FRACTIONS WITH THRESHOLD = *E15.7
1 * RADIANS.*73X*NONLINEAR -*E15.7*, PHASE-LOCKED LOC -*E15.7//)
XMCN=XMCCSV+CSN
XMCPL=XMCPL+GSP
XMCN=0.
XMCPP=0.
XMCN=0.
XMCPP=0.
XMCVP=J.
H=N02-3J
DO 6300 I=31,N02
XMCN=XMCN+XDAT(I,1)
XMCPP=XMCPP+XDAT(I,3)
630. XMCVP=XMCVP+XDAT(I,5)
XMCN=XMCN/4
XMCPP=XMCPP/4
XMCVP=XMCVP/H
SHN=XMCN*XMCN
SMP=XMCPP*XMCPP
DO 6400 I=31,N02
XMCN=XMCN+XDAT(I,2)

```

```

PROGRAM          MAIN
390              5406  XMCCP=XMCCP+XDATH(I,4)
                  XMCCN=XMCCN/H-SMN
                  XMCCP=XMCCP/H-SMP
                  ALXN=16.*AL0510(XMCCN)
                  ALXP=10.*AL0510(XMCCP)
                  WRITE (6,943.2) NMC,XMCCN,XMCCP,XMCCP,XMCCP,ALXN,ALXP
395              9036J  FORMAT('MONTE CARLO NO.*15,10X*ENLF*12X*VNLF*12X*EPLL*12X
                  1 *VPLL*12X*CVPLL*/' * CUMULATIVE TIME AVERAGE*5E15.7/' * DECIBELS*
                  2 31X,E15.7,13X,E15.7)
                  NMC=NMC+1
                  IF (NMC.GT.JSEED) STOP
                  GO TO 9000
                  END

```

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SUMMARY OF CHANGES MADE BY THE OPTIMIZER

```

27 WORDS OF INVARIANT RLIST REMOVED FROM THE LOOP STARTING AT LINE 121
13 WORDS OF INVARIANT RLIST REMOVED FROM THE LOOP STARTING AT LINE 126
28 WORDS OF INVARIANT RLIST REMOVED FROM THE LOOP STARTING AT LINE 137
14 WORDS OF INVARIANT RLIST REMOVED FROM THE LOOP STARTING AT LINE 204

```

REGISTER ALLOCATION

```

4 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 92
4 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 92
4 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 92
1 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 106
1 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 126
3 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 142
3 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 220
3 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 229
3 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 236
3 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 245
1 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 252
2 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 269
3 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 275
2 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 375
2 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 384

```

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PROGRAM MAIN

SYMBOLIC REFERENCE MAP

ENTRY POINTS JEF LINE REFERENCES

1

VARIABLES SN TYPE RELOCATION

4052 MAIN	6173 A	REAL	84	DEFINED	83		
	6210 AI	REAL	150	DEFINED	147		
	6211 AJ	REAL	150	DEFINED	149		
	6212 AK	REAL	151	DEFINED	155	150	
	6125 ALP110	REAL	26	33	35	20	
	6265 ALXN	REAL	391	DEFINED	389		
	6266 ALXP	REAL	391	DEFINED	390		
	33164 AN	REAL	12				
	6143 A11	REAL	73	138	163	33	
	6144 A22	REAL	74	138	164	34	
	6241 CHAT	REAL	277	279	280	DEFINED	273
	6245 COSF1	REAL	293	294	292	277	277
	21002 COSY	REAL	7	205	277	114	
	6230 CS	REAL	178	179	312	343	364
		REAL	175	337	312		
	6250 CSN	REAL	345	364	367	338	345
	6251 CSP	REAL	352	364	368	339	352
	6242 CXHAT	REAL	281	282	284	279	282
	6126 DOLF	REAL	30	DEFINED	20		
	6157 DELSQ	REAL	2*192	DEFINED	59		83
	6140 DELT	REAL	31	32	35	62	
		REAL	192	2*194	336	30	
	6172 DEN	REAL	192	194	195	287	293
		REAL	294	78	200		
	0 DENS1	REAL	2	139	143	2*224	2*240
		REAL	2*252	276	277	DEFINED	143
		REAL	270	265			
	4235 DENS2	REAL	2	DEFINED	171		
	6225 DEVQ2	REAL	188	DEFINED	163		
	6216 DEV1	REAL	168	DEFINED	164		
	6217 DEV4	REAL	169	170	189	DEFINED	165
	6220 DEV3	REAL	224	233	240	256	261
	6236 DUH	REAL	264	265	217	233	249
		REAL	251	263			
	0 OZZZ1	REAL	5				
	6176 EOG1	REAL	95	DEFINED	90		
	6177 EOG2	REAL	99	DEFINED	91		
	33202 EIGVEC	REAL	12				
	6231 ENL	REAL	178	312	176	313	
	6232 EPL	REAL	179	313	177	311	
	6252 ERRN	REAL	341	343	2*347	363	
	6254 ERRP	REAL	348	350	2*354	364	
	21161 EXDON	REAL	7	125	128	DEFINED	125
	21123 EXPON	REAL	7	133	233	240	249
		REAL	127	128	131		
	33166 F	REAL	12	DEFINED	83	81	82
	6137 FTC	REAL	53	2*34	35	29	

PROGRAM			MAIN		RELOCATION		CDC 660C FTN V3.0-P213 OPT=2										03/22/72 16.47.23.		PAGE		10	
VARIABLES	SN	TYPE					REFS	379	380	381	387	388										
C264 H		REAL					DEFINED	374														
6154 I		INTEGER					REFS	55	93	95	97	99										
							106	108	2*114	2*115	120	3*125										
							133	2*138	139	2*143	147	155										
							158	159	160	161	181	3*183										
							223	231	232	238	239	247										
							259	261	265	2*270	2*276	2*277										
							2*341	2*346	2*347	2*348	2*353	2*354										
							357	358	359	4*360	376	377										
							386	DEFINED	53	92	96	100										
							113	119	126	133	136	141										
							204	215	268	274	314	346										
6174 IA		INTEGER					REFS	86	117	122	DEFINED	84										
6201 IB		INTEGER					REFS	121	DEFINED	117												
6203 IC		INTEGER					REFS	123	DEFINED	122												
6253 ICSNLF		INTEGER					REFS	363	DEFINED	342	344											
6255 ICSPLL		INTEGER					REFS	350	DEFINED	349	351											
6202 IR		INTEGER					REFS	122	DEFINED	121												
6204 IYY		INTEGER					REFS	133	DEFINED	132												
6175 IY2		INTEGER					REFS	86	100	118	119	126										
							131	132	209	210	DEFINED	85										
6155 J		INTEGER					REFS	55	2*138	139	DEFINED	149										
							157	158	159	160	161	183										
							265	2*270	276	277	317	DEFINED										
							142	148	182	206	269	275										
1 JGAUSS		INTEGER	LN				REFS	5	DEFINED	19												
6133 JSECO		INTEGER					REFS	22	24	58	166	168										
6213 KL		INTEGER					REFS	189	190	396	DEFINED	20										
							REFS	152	153	154	156	2*157										
							3*233	3*240	3*249	3*256	DEFINED	151										
							231	238	247	254												
11472 KLOM		INTEGER	ARRAY		/ /		REFS	2	222	231	238	247										
							DEFINED	156	157	158												
6156 KOUNT		INTEGER					REFS	167	187	284	285	308										
							2*311	328	329	DEFINED	56	329										
6214 KU		INTEGER					REFS	159	2*160	2*161	224	233										
							256	DEFINED	153	223	232	239										
14727 KUP		INTEGER	ARRAY		/ /		REFS	2	223	232	239	248										
							DEFINED	159	160	161												
6233 K1		INTEGER					REFS	211	213	220	229	245										
6234 K2		INTEGER					REFS	212	214	220	236	252										
6237 L		INTEGER					REFS	222	223	220	236	252										
							239	4*240	247	4*224	231	232										
6117 LOW		LOGICAL					REFS	220	229	236	245	252										
6235 MM		INTEGER					REFS	15	213	218	207	211										
							REFS	221	224	230	233	237										
							249	253	256	DEFINED	216	221										
							246	253														
5725 NAS		INTEGER					REFS	344	351	DEFINED	17	346										
5724 NBL		INTEGER					REFS	342	349	DEFINED	17											
6153 NMC		INTEGER					REFS	58	86	104	333	347										

PROGRAM	MAIN	RE-LOCATION	CDC 5500 FIN V3.0-P213 OPT=2	03/22/72	15.47.23.	PAGE	11
VARIABLES	SN	TYPE					
6132	NO2	INTEGER	354 REFS	355 53 DEFINED	391 328 20	395 340 DEFINED	52 374 395 375
6130	NUM1	INTEGER	384 REFS	65 2*158 157 274 314 DEFINED	92 160 20 108 214 20	113 180 204 136 141 215 146 268	
6131	NUM2	INTEGER	354 REFS	206 212 213 20 DEFINED	96 213 312 313 178 179 174	137 229 142 245 178 179 192 194	
6226	NUNNL	INTEGER	316 REFS	316 REFS	20 312	173 174	312 313
6227	NUNPL	INTEGER	354 REFS	178 REFS	312 313	173 174	312 313
33154	PBAR	REAL	354 REFS	178 REFS	312 313	173 174	312 313
33172	POUNY	REAL	354 REFS	178 REFS	312 313	173 174	312 313
33176	POUNY2	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6160	PI	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6162	PIOLT	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6163	PIINV	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6161	PI2	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6164	PI2OLT	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6243	PLOSS	REAL	354 REFS	178 REFS	312 313	173 174	312 313
33166	PN	REAL	354 REFS	178 REFS	312 313	173 174	312 313
33206	PNF	REAL	354 REFS	178 REFS	312 313	173 174	312 313
21164	PROP	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6134	P115	REAL	354 REFS	178 REFS	312 313	173 174	312 313
33150	Q	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6135	QQ	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6141	Q22	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6127	Q22C	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6171	R	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6207	RATIO	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6215	REALK	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6136	RA	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6142	R11	REAL	354 REFS	178 REFS	312 313	173 174	312 313
33027	SENS	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6240	SHAT	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6244	SINF1	REAL	354 REFS	178 REFS	312 313	173 174	312 313
20661	SINY	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6256	SMN	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6257	SMP	REAL	354 REFS	178 REFS	312 313	173 174	312 313
6122	SPL	LOGICAL	354 REFS	178 REFS	312 313	173 174	312 313
6121	SPLST	LOGICAL	354 REFS	178 REFS	312 313	173 174	312 313

PROGRAM			MAIN		CDC 56J- FTN V3.0-P213 OPT=2 03/22/72 16.47.23.										PAGE	12
VARIABLES	SN	TYPE	RELOCATION		REFS	139	140	264	267	DEFINED	135	139				
6205 SUM		REAL			203	264										
6206 SUMI		REAL			REFS	143	270	DEFINED	143	267						
6120 UP		LOGICAL			REFS	15	214	219	DEFINED	208	212					
6165 U1		REAL			REFS	90	94	145	150	DEFINED	65					
6166 U2		REAL			REFS	85	91	98	102	145						
					DEFINED	66										
6200 X		REAL			REFS	94	95	98	99	102	103					
					DEFINED	93	94	97	98	101	102					
6267 XDAT		REAL	ARRAY		REFS	6	2*176	2*177	2*310	2*311	2*341					
					DEFINED	167	167	187	284	285	308					2*348
21200 XDATH		REAL	ARRAY		REFS	9	346	347	353	354	355					309
					REFS	357	359	3*360	376	377	378					356
					REFS	386	55	346	347	353	354					385
					DEFINED	14	183	317	DEFINED	181	183					355
33212 XDEN		REAL	ARRAY		REFS	317										315
					REFS	350	DEFINED	358								
6260 XOMSN		REAL			REFS	360	DEFINED	359								
6261 XOMSP		REAL			REFS	6	295	2*306	307	308						
20077 XHAT		REAL	ARRAY		DEFINED	69	70	293	294	71	306					
					REFS	291	292	293	DEFINED							
6167 XHAT1		REAL			REFS	294	DEFINED	72	307	DEFINED	47					371
6170 XHAT2		REAL			REFS	385	387	389	391	DEFINED						
6146 XMCN		REAL			REFS	385										
					REFS	387										
6151 XMCOP		REAL			REFS	386	388	390	391	DEFINED	50					372
					REFS	388										
6147 XMCOSN		REAL			REFS	357	DEFINED	48	367							
6152 XMCOSP		REAL			DEFINED	51										
6263 XMCVCP		REAL			REFS	378	381	391	DEFINED	373	378					381
6262 XMCPLL		REAL			REFS	368	DEFINED	368								
6145 XMCN		REAL			REFS	376	379	2*382	391	DEFINED	46					369
					REFS	376										
6150 XMCPP		REAL			REFS	377	380	2*383	391	DEFINED	49					370
					REFS	380										
3 XNZ		REAL	ARRAY	RN	REFS	4										
2 XZZZ		REAL	ARRAY	GN	REFS	5										
6221 X1		REAL			REFS	166	167	169	170	186	187					189
					REFS	318	DEFINED	186								
6246 X1FF		REAL			REFS	297	298	300	302	DEFINED	295					340
					REFS	302										
6247 X1MOD		REAL			REFS	320	321	323	325	DEFINED	318					323
					REFS	168	186	2*188								
6222 X2		REAL			REFS	7	106	114	115	138						
20101 Y1		REAL	ARRAY		DEFINED	95										
					REFS	69	71	166	DEFINED	20						
6123 Y1EST		REAL			REFS	70	108	138	DEFINED	99						
20222 Y2		REAL	ARRAY		REFS	70	72	168	DEFINED	20						
6124 Y2EST		REAL			REFS	10	125	168	DEFINED	103						
33010 Y3		REAL	ARRAY		REFS	169	189	205	293	294						
6223 Z1		REAL			REFS	170	190	205	293	294						
6224 Z2		REAL			REFS			205	293	294						

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LOOPS	LABEL	INDEX	FROM-TO	LENGTH	PROPERTIES
5122	307	L	252 258	138	NOT INNER
5161	312	* I	268 271	128	INSTACK
5167	312	J	269 271	28	NOT INNER
5176	360	* I	274 278	168	INSTACK
5205	360	J	275 278	38	NOT INNER
5324	5676	* I	314 317	138	INSTACK
5332	5676	J	316 317	28	EXT REFS
5373	6200	* I	340 363	1228	INSTACK
5545	6300	I	375 378	48	INSTACK
5565	6400	I	384 386	38	INSTACK

COMMON BLOCKS	LENGTH	MEMBERS - BIAS NAME(LENGTH)
/ /	11325	3 DENS1 (2205)
RN	5	6615 KU2 (2205)
GN	4	3 NZZZ (3)
		3 DZZZ1 (1)

STATISTICS	PROGRAM LENGTH	271758
PROGRAM LENGTH	271758	119J1
BUFFER LENGTH	40448	2084
COMMON LENGTH	118	9
BLANK COMMON	254218	11025

2205 DENS2 (2205)	4410 KLOW (2205)
9820 PROP (2205)	
3 ANZ (2)	
1 JGAUSS (1)	2 XZZZ (2)

SUBROUTINE MOD2PI

```
5      SUBROUTINE MOD2PI(X,XM)
        DATA ISK/0/
        IF (ISK) 20, 10, 20
10      PI=4.*ATAN(1.)
        IWOPI=2.*PI
        ISM=999
        XM=X
20      XM=X
30      IF (XM.GT.PI) GO TO 40
        IF (XM.LT.-PI) GO TO 50
        RETURN
40      XM=XM-IWOPI
        GO TO 30
50      XM=XM+IWOPI
        GO TO 30
        END
```

CDC 660C FTN V3.0-P213 OPT=2 03/22/72 16.47.23.

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1

FUNCTION AVG

FUNCTION AVG(AV,K,I)
AVG=((I-1.)*AV+A)/I
RETURN
END

CDC 660C FIN V.0.0-P213 OPT=2 03/22/72 16.47.23.

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FUNCTION		AVG	
SYMBOLIC REFERENCE MAP			
ENTRY POINTS	DEF LINE	REFERENCES	
2 AVG	1	3	
VARIABLES	SN	TYPE	RELOCATION
0 AV		REAL	F.P.
14 AVG		REAL	
0 I		INTEGER	F.P.
0 X		REAL	F.P.
STATISTICS			
PROGRAM LENGTH	158	13	

REFS	2	DEFINED	1
DEFINED	2		
REFS	2*2	DEFINED	1
REFS	2	DEFINED	

```

5      FUNCTION RANF(NS,MODE)
        DIMENSION NS(2), NC(2)
        COMMON /RNF/ N1, N2, MP, I1, I2
        DATA N1, N2/244734, 158551/
        MODE=0 TO CONTINUE, OTHERWISE RESTART WITH
        INTEGER NUMBER NS(1)*2**18+NS(2)
        IF (MODE) 10, 100, 10
10      N1=NS(1)
        N2=NS(2)
        I1=2.**(-18)
        I2=2.**(-36)
        MP=2.**18
100     DO 200 J=1,2
        GO TO (110,120),I
110     K=M2*N2
        GO TO 190
120     K=M1*N2+M2*N1+KD
190     KD=K/MP
200     NC(I)=K-KD*MP
        N1=NC(2)
        N2=NC(1)
        XN1=N1
        XN2=N2
        RANF=XN1*I1+XN2*I2
        RETURN
        END

```

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COC 6600 FIN V3.0-P213 OPT=2 03/22/72 16.47.23.

FUNCTION RANF

SYMBOLIC REFERENCE MAP

ENTRY POINTS	DEF LINE	REFERENCES
2 RANF	1	25

VARIABLES	SN	TYPE	RELOCATION
76 I		INTEGER	
77 K		INTEGER	
100 KD		INTEGER	
6 MODE		INTEGER	
2 MP		INTEGER	
72 M1		INTEGER	
73 M2		INTEGER	
103 NC		INTEGER	
0 NS		INTEGER	
0 N1		INTEGER	
1 N2		INTEGER	
75 RANF		REAL	
3 T1		REAL	
4 T2		REAL	
101 XN1		REAL	
102 XN2		REAL	

REFS	DEF LINE	REFERENCES
14	19	DEFINED
18	19	DEFINED
17	19	DEFINED
7	0	DEFINED
3	18	DEFINED
17	17	DEFINED
15	17	DEFINED
2	20	DEFINED
2	8	DEFINED
3	17	DEFINED
3	15	DEFINED
24	24	DEFINED
3	24	DEFINED
3	24	DEFINED
24	24	DEFINED
24	24	DEFINED
24	24	DEFINED

STATEMENT LABELS	DEF LINE	REFERENCES
0 10	8	2*7
27 100	13	7
41 110	15	14
44 120	17	14
51 190	18	16
0 200	19	13

LOOPS LABEL	INDEX	FROM-TO	LENGTH	PROPERTIES
33 200	1	13 19	258	OPT

COMMON BLOCKS	LENGTH	MEMBERS	BIAS NAME(LENGTH)
RN	5	0 N1	(1)
		3 T1	(1)

STATISTICS	PROGRAM LENGTH	COMMON LENGTH
	1128	58

1 N2	4 T2	2 MP	(1)
1	4	2	(1)

SUBROUTINE GAUSS

COC 6600 FTN V3.0-P213 OFI=2 03/22/72 16.47.23.

PAGE 1

```
5      SUBROUTINE GAUSS(JS,SD,XM,X)
        DIMENSION NST(2)
        COMMON /RN/ N1, N2, HC, I1, I2
        COMMON /GN/ TWOPI, J, XR(2)
        IF (J) 10, 15, 20
10      J=2
        TWOPI=8.*ATAN(1.)
        NST(1)=24734
        NST(2)=158551
15      NST(1)=102943
        NST(2)=185617
        XR(1)=RANF(NST,1)
        GO TO 35
25      GO TO (30,40), J
30      J=2
        XR(1)=RANF(NST,J)
        XR(2)=RANF(NST,J)
        X1=SQRT(ABS(-2.*ALOG(XR(1))))
        XR(2)=TWOPI*XR(2)
        XR(1)=X1*SIN(XR(2))
        XR(2)=X1*COS(XR(2))
        X=XR(1)*SD+XM
        RETURN
40      J=1
        X=XR(2)*SD+XM
        RETURN
        END
```


SYMBOLIC REFERENCE MAP

VARIABLES	SN	TYPE	RELJCATN
1	1	1	1
2	2	2	2
3	3	3	3
4	4	4	4
5	5	5	5
6	6	6	6
7	7	7	7
8	8	8	8
9	9	9	9
10	10	10	10
11	11	11	11
12	12	12	12
13	13	13	13
14	14	14	14
15	15	15	15
16	16	16	16
17	17	17	17
18	18	18	18
19	19	19	19
20	20	20	20
21	21	21	21
22	22	22	22
23	23	23	23
24	24	24	24
25	25	25	25
26	26	26	26
27	27	27	27
28	28	28	28
29	29	29	29
30	30	30	30
31	31	31	31
32	32	32	32
33	33	33	33
34	34	34	34
35	35	35	35
36	36	36	36
37	37	37	37
38	38	38	38
39	39	39	39
40	40	40	40
41	41	41	41
42	42	42	42
43	43	43	43
44	44	44	44
45	45	45	45
46	46	46	46
47	47	47	47
48	48	48	48
49	49	49	49
50	50	50	50
51	51	51	51
52	52	52	52
53	53	53	53
54	54	54	54
55	55	55	55
56	56	56	56
57	57	57	57
58	58	58	58
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61	61	61	61
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67	67	67	67
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74	74	74	74
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100	100	100	100

1	J	INTEGER	GN
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```
G JS      *UNUSED   F.P.  
INTEGER  
INTEGER
```

THE

[illegible]

3	T1	REAL	RN
3	T1	REAL	RN

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EXTERNALS	TYPE	APCS	REFERENCE NO

ATAN	REAL	1	LIBRARY	7
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RANF		REAL	Z	1.79893E-06
SIN		REAL	Z	-1.79893E-06

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0	10	6	2*	

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UNCLASSIFIED

PROGRAM	LENGTH	COMMON	LENGTH
1.28	1.28	1.18	9
68	68	68	68

1.0435E-07 3.8767E-13

Additional Appendix E.

A Fourier Series Implementation
of the Cyclic Phase Demodulator

The Fourier Series experiment, described in Appendix D of Chapter VII is included here.

The program was compiled on one of the Kirtland AFB CDC 6600's using the FTN compiler. The code-optimization algorithm was employed. The random number generator is identical to the one described in Chapter IV.

Programmer: C. Hecht

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PROGRAM MAIN

```

        AJ=J**2
        ARG=EXP(Q*EXP(AJ))
        EMSL(INTERM2+J)=ARG
        EMSL(INTERM2-J)=ARG
451      CONTINUE
        EMSL(INTERM2)=1.0
        PC1=-.5*A11
        PC2=-.5*A22
        DO 452 I=1,INTERM1
          DO 452 J=1,INTERM1
            AT=(I-INTERM2)**2
            AJ=(J-INTERM2)**2
            XAA=EXP(PC1*AI + PC2*AJ)
            AA1(I,J)=CMPLX(XAA,0.0)
            AA2(I,J)=( ,1.)
452      CONTINUE
            Q(1,1)=J.0
            Q(2,2)=Q22
            R=0.11
75      2 CONTINUE
            KOUNT=1
            XHAT(1)=Y1EST
            XHAT(2)=Y2EST
            XHAT1=Y1EST
            XHAT2=Y2EST
            PH(1,1)=A11
            PH(2,2)=A22
            PH(1,2)=.
            PH(2,1)=.
            DELV=1.0/(PN(1,1) + R)
            F(1,1)=1.0
            F(1,2)=DELT
            F(2,1)=0.
            F(2,2)=1.0
            DEV1= SQR(A11)
            DEV2= SQR(A22)
            DEV3= SQR(R)
79      CALL GAUSS(JSEED,DEV1,Y1EST,X(1))
            CALL GAUSS(JSEED,DEV2,Y2EST,X(2))
            XDAT(KOUNT,1)=X(1)
            CALL GAUSS(JSEED,DEV3,COS(X(1)),Z1)
            CALL GAUSS(JSEED,DEV3,SIN(X(1)),Z2)
            DEV01= SQR(Q(1,1))
            DEV02= SQR(Q(2,2))
            GO TO 78
100      CONTINUE
            X(1)=X(1) + X(2)*DELT
            XDAT(KOUNT,1)=X(1)
            CALL GAUSS(JSEED,DEV2,X(2),X(2))
            CALL GAUSS(JSEED,DEV3,COS(X(1)),Z1)
            CALL GAUSS(JSEED,DEV3,SIN(X(1)),Z2)
            PICCATI=EQUATION UPDATE *****
            PDIMY(1,1)=(R*(PN(1,1)+2.6*PH(1,2)*DELT)-PN(1,2)*DELT)*JEN
            PDIMY(1,2)=PH(1,2)*(R-PN(1,2)*DELT)*DEN + PN(2,2)*DELT
110

```

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PDJHY(2,2)=-PN(1,2)**2*DEN + PN(2,2) + 0(2,2)

PN(1,1)=PDJHY(1,1)

PN(1,2)=PDJHY(1,2)

PN(2,2)=PDJHY(2,2)

PI(2,1) = PN(1,2)

DN=1.0/(PN(1,1) + R)

7 CONTINUE

09MAG=SQR(IZ1**2 + Z2**2)/R

03PH=ATAN2(Z2,Z1)

CALL IL(08MAG,BSF)

CALL INUL(08MAG,NT,RM,BSF,BSF2)

00 53 K=1,INTER1

AK=K

CMX3=CMPLX(0.0,08PH*AK)

CMXL=CMPLX(0.0,08PH*AK)

SSK(INTERM1-K)=C4PLX(3SF2(K),...) *C4XP(CMPX3)

SSK(K+INTERM1)=C4PLX(3SF2(K),...) *C4XP(CMPX4)

530 CONTINUE

SSK(INTERM1)=C4PLX(BSF,U.)

CMX2=(0.0,0.0)

00 54 NU=1,INTERM1

CMX=SSK(-NU+INTERM1)*AA1(NU,NU)

CMX2=CMX2+C4PX

540 CONTINUE

CMX3=CMX2

00 55 L=1,INTERM1

CMX4=CMPLX(0.0,MSL(L)...)/CMX3

00 56 L=1,INTERM1

CMX2=(0.0,0.0)

00 55 NU=1,INTERM1

IND=L+NU-INTERM2

IF (IND.LT.2.OR.IND.GT.INTERM1) GO TO 55

CMX=SSK(M-NU+INTERM1)*AA1(NU,IND)

CMX2=CMX2+C4PX

550 CONTINUE

AA2(M,L)=CMX2+CMX2

56 CONTINUE

4122 FORMAT(*,IP1,11.3)

00 57 M=1,INTERM1

00 57 L=1,INTERM1

AA1(M,L)=AA2(M,L)

57 CONTINUE

M1=INTERM2-1

N=INTERM2

AA11=AIMAG(AA1(N1,M1))

AA22=REAL(AA1(N1,N))

X1NL=ATAN2(AA11,AA22)

XUAT(KOUNT,3)=X1NL

PNF(1,1)=PN(1,1)*R*O_N

PNF(1,2)=PN(1,2)*R*O_N

PNF(2,1)=PNF(1,2)

PNF(2,2)=PNF(2,2) - PN(1,2)**2*DEN

C ***

FILT,R U,JAfc ****

SIMF1=SIN(XHAT1)

CO:FI=COS(XHAT1)

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PROGRAM MAIN

STOP
END

CDC 6600 FTN V3.0-P213 OPT=2 03/26/72 01.12.49.

PAGE

SUMMARY OF CHANGES MADE BY THE OPTIMIZER
14 WORDS OF INVARIANT RLIST REMOVED FROM THE LOOP STARTING AT LINE 55

SYMBOLIC REFERENCE MAP

ENTRY POINTS	DEF LINE	REFERENCES	
4052 MAIN	1		
VARIABLES	SN	TYPE	RELOCATION
2545 AA1		COMPLEX	ARRAY
2206 AA11		REAL	
14207 AA2		COMPLEX	ARRAY
2267 AA22		REAL	
5235 AI		REAL	
5230 AJ		REAL	
2257 AK		REAL	
5167 ALP110		REAL	
23761 AN		REAL	*UNDEF
5231 AR6		REAL	
5211 A11		REAL	
5212 A22		REAL	
5255 BSF		REAL	
22651 BSF2		REAL	
5195 CMX		COMPLEX	ARRAY
5157 CMX2		COMPLEX	
5161 CMX3		COMPLEX	
5163 CMX4		COMPLEX	
5272 COSF1		REAL	
5170 DELF		REAL	
5225 DELSN		REAL	
5205 DELT		REAL	
5243 DEN		REAL	
2251 DEVOL	*	REAL	
2252 DEVQ2		REAL	
5244 DEV1		REAL	
5245 DEV2		REAL	
2246 DEV3		REAL	
5277 JUMP1		REAL	
5300 USUMP2		REAL	
6 UZZ1		REAL	
24031 EIGVEC		REAL	
23015 ENSL		REAL	
23763 F		REAL	
2234 FTC		REAL	
5215 H		REAL	
5234 I		INTEGER	
5263 IMJ		INTEGER	
5213 ISAMP		INTEGER	
5227 J		INTEGER	
1 JGAUSS		INTEGER	
75 JS-EU		INTEGER	

REFS	2	132	143	155	156	
DEFINED	69	151				
REFS	157	0-DEFINED	155	70	146	
REFS	2	151	0-DEFINED			
REFS	157	0-DEFINED	156			
REFS	68	68	68	56	67	
REFS	57	68	0-DEFINED	123		
REFS	124	125	0-DEFINED	12		
REFS	15	27	0-DEFINED			
REFS	6	59	0-DEFINED	37		
REFS	58	62	81	90	DEFINED	26
REFS	31	63	82	91	DEFINED	28
REFS	12	121	129			
REFS	4	121	126	127		
REFS	3	133	144	0-DEFINED	132	143
REFS	3	133	135	144	146	
REFS	130	133	139	144		
DEFINED	130	126	137	0-DEFINED	124	135
REFS	3	127	146	0-DEFINED	125	137
REFS	3	167	0-DEFINED	165		
REFS	166	0-DEFINED	12			
REFS	22	2*108	0-DEFINED	33		
REFS	54	24	29	53	87	102
REFS	23	24	22			
REFS	169	0-DEFINED	111	159	162	166
REFS	103	110	116			
REFS	167	85				
DEFINED	93					
REFS	104	0-DEFINED	99			
REFS	93	0-DEFINED	91			
REFS	94	0-DEFINED	91			
REFS	96	97	105	105	DEFINED	92
REFS	211	0-DEFINED	196			
REFS	211	0-DEFINED	210			
REFS	9					
REFS	6	137	0-DEFINED	58	59	61
REFS	6	0-DEFINED	86	87	88	89
REFS	22	29	0-DEFINED	21		
REFS	191	208	0-DEFINED	47		
REFS	66	69	70	2*182	2*199	
REFS	64	181	198			
DEFINED	64	143	0-DEFINED	141		
REFS	2*142	215	216	0-DEFINED	42	216
REFS	211	58	59	67	69	71
REFS	56	58				
REFS	55	65				
0-DEFINED	9	0-DEFINED	11	97	104	105
REFS	93	94	96			106

VARIABLES	SV	TYPE	RELOCATION	DEFINED	12	2*126	2*127	DEFINED	122	176	177	452
2256 K	INTEGER			REFS	123	2*126	2*127	DEFINED	122	176	177	
5240 KOUNT	INTEGER			REFS	95	103	158	171	2*172			
2261 L	INTEGER			REFS	198	DEFINED	76	177	DEFINED	136	150	
2262 M	INTEGER			REFS	137	141	146	2*151	DEFINED	138		
5173 M01	INTEGER			REFS	143	146	2*151	DEFINED	138	149		
5174 M02	INTEGER			REFS	48	50	52	55	DEFINED	12		
5214 NSAMP	INTEGER			REFS	44	172	176	DEFINED	12			
5223 NTC	INTEGER			REFS	193	214	DEFINED	43	214			
5220 NTERM	INTEGER			REFS	132	DEFINED	51	DEFINED	48	129	131	
5221 NTERM1	INTEGER			REFS	49	121	122	126	127	150		
				REFS	51	64	65	143	149			
				REFS	138	140	142					
				REFS	136							
2222 NTERM2	INTEGER			DEFINED	49	58	59	61	66	67	141	
				REFS	51	DEFINED	50					
2224 NTERM3	INTEGER			REFS	154							
2260 NU	INTEGER			DEFINED	52							
2172 NUH1	INTEGER			REFS	3*132	141	2*143	DEFINED	131	142		
2172 NUH2	INTEGER			DEFINED	12							
5265 N1	INTEGER			REFS	8							
5266 N1	INTEGER			REFS	155	156	DEFINED	154				
2253 OBHAG	REAL			REFS	155	156	DEFINED	153				
2254 OBPH	REAL			REFS	124	125	DEFINED	118				
23751 PUAR	REAL			REFS	6			119				
5232 PC1	REAL			REFS	68	DEFINED	52					
5233 PC2	REAL			REFS	68	DEFINED	63					
5231 P08	REAL			REFS	29	DEFINED	18					
23767 POUHY	REAL			REFS	6	112	113	114	DEFINED	108	119	
				REFS	111							
23773 POUHY2	REAL			REFS	6							
5177 PI	REAL			REFS	17	184	201	DEFINED	16			
5200 PI2	REAL			REFS	186	203	DEFINED	17				
23755 PN	REAL			REFS	6	85	*138	3*110	2*111	115	116	
				REFS	163	2*162	2*165	2*167	DEFINED	81	82	
				REFS	159	112	113	114	115			
				REFS	83	161	DEFINED	159	161	162		
24037 PNF	REAL			REFS	6	29	25	26	29			
5176 P1.0	REAL			REFS	18							
				DEFINED	15							
5210 P22J	REAL			REFS	28	DEFINED	25	111	DEFINED	72	73	
23745 Q	REAL			REFS	6	98	99					
2266 QEXP	REAL			REFS	57	DEFINED	54					
2202 QQ	REAL			REFS	20	21	DEFINED	19				
5216 Q22	REAL			REFS	19	54	73	DEFINED	23			
5171 Q22C	REAL			REFS	29	23	25	30	DEFINED	12	159	
5237 R	REAL			REFS	85	92	108	111	116			
				REFS	163	74						
5275 RUMP	REAL			DEFINED	211	DEFINED	192		DEFINED	20		
5203 RX	REAL			REFS	21	24	25	30				
5207 R11	REAL			REFS	29	74	DEFINED	24				
5271 SINF1	REAL			REFS	166	167	DEFINED	164				
5301 SSK	COMPLEX			REFS	2	132	1-3	DEFINED	126	127	129	
5273 SUMP	REAL			REFS	189	191	192	193	206	208	209	
				REFS	211	180	189	191	197			
				DEFINED								

PROGRAM MAIN

STATEMENT LABELS

DEF LINE REFERENCES

0 452 71 61 63

0 230 120 122

0 240 134 131

4542 250 141 142

0 260 147 136

0 270 152 149

5101 021 35 12

5064 052 36 29

5147 053 33 30

5140 054 32 31

5114 050 17 172

5127 051 212 211

0 1601 19 181

4673 1898 183 187

4675 1699 130 184

4677 1710 188 185

0 1731 14 13

4770 1799 22 13

4767 1800 218 210

0 2601 207 198

4722 2698 20 204

4724 2699 213 201

4726 2710 210 202

5111 4122 148

INACTIVE

PROPS

LOOPS LABEL FROM-TO LENGTH PROPERTIES
 4225 451 * J 61 128
 4246 452 * I 64 71 278
 4251 452 * J 71 228
 4421 451 * K 128 3 8
 4464 445 * NU 131 134 138
 4543 460 * L 136 147 508
 4513 460 * M 138 147 478
 4521 451 * NU 140 238
 4562 451 * M 149 148
 4571 451 * L 150 38
 4671 1601 * I 181 118
 4720 2601 * I 198 118

EXT REFS NOT INNER
 EXT REFS NOT INNER
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COMMON BLOCKS

RN LENGTH MEMBERS - BIAS NAME (LENGTH)
 5 0 4222 (3)
 4 3 JZZZ1 (1)

3 XNZ (2)
 1 JGAUSS (1)
 2 XZZZ (2)

STATISTICS

PROGRAM LENGTH 177518 8169
 BUFFER LENGTH 40469 2184
 COMMON LENGTH 118 9

ENTRY POINTS	2	10	DEF LINE	33
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REFERENCES
39 44

VARIABLES	SN	TYPE
Q R10		REAL

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.....
SUBROUTINE INUE
PURPOSE
COMPUTE THE MODIFIED BESSEL FUNCTIONS I FOR ORDERS 1 TO N
USAGE
CALL INUE(X,N,ZI,RI)
DESCRIPTION OF PARAMETERS
X - GIVEN ARGUMENT OF THE BESSEL FUNCTIONS I
N - GIVEN MAXIMUM ORDER OF BESSEL FUNCTIONS I
ZI - GIVEN VALUE OF BESSEL FUNCTION I OF ORDER ZERO
FOR ARGUMENT X
RI - RESULTANT VECTOR OF DIMENSION N, CONTAINING THE
VALUES OF THE FUNCTIONS I FOR ORDERS 1 TO N
REMARKS
THE VALUE OF ZI MAY BE CALCULATED USING SUBROUTINE IO.
USING A DIFFERENT VALUE HAS THE EFFECT THAT ALL VALUES OF
BESSEL FUNCTIONS I ARE MULTIPLIED BY THE FACTOR ZI/(G,X)
WHERE I(G,X) IS THE VALUE OF I FOR ORDER G AND ARGUMENT X.
THIS MAY BE USED DISADVANTAGEOUSLY IF ONLY THE RATIOS OF I
FOR DIFFERENT ORDERS ARE REQUIRED.
SUBROUTINES AND FUNCTION SUBPROGRAMS REQUIRED
NONE
METHOD
THE VALUES ARE OBTAINED USING BACKWARD RECURRENCE RELATION
TECHNIQUE. THE RATIO I(N+1,X)/I(N,X) IS OBTAINED FROM A
CONTINUED FRACTION.
FOR REFERENCE SEE
G. BLANCH, NUMERICAL EVALUATION OF CONTINUED FRACTIONS*,
SIAM REVIEW, VOL.6, NO.1, 1964, PP.383-421.
.....
SUBROUTINE INUE(X,N,ZI,RI)
DIMENSION RI(1)
IF(N) GO TO 1
1 EN=N
2 AN=1.
3 FI=FI+2.
4 AN=FI/ABS(X)
5 A=AN*A1+A0
6 B=4N*B1+B0
7 FI=FI
8 AN=FI+2.
9 AN=FI/ABS(X)
10 A=AN*A1+A0
11 B=4N*B1+B0
12 FI=FI
13 AN=FI+2.
14 AN=FI/ABS(X)
15 A=AN*A1+A0
16 B=4N*B1+B0
17 FI=FI
18 AN=FI+2.
19 AN=FI/ABS(X)
20 A=AN*A1+A0
21 B=4N*B1+B0
22 FI=FI
23 AN=FI+2.
24 AN=FI/ABS(X)
25 A=AN*A1+A0
26 B=4N*B1+B0
27 FI=FI
28 AN=FI+2.
29 AN=FI/ABS(X)
30 A=AN*A1+A0
31 B=4N*B1+B0
32 FI=FI
33 AN=FI+2.
34 AN=FI/ABS(X)
35 A=AN*A1+A0
36 B=4N*B1+B0
37 FI=FI
38 AN=FI+2.
39 AN=FI/ABS(X)
40 A=AN*A1+A0
41 B=4N*B1+B0
42 FI=FI
43 AN=FI+2.
44 AN=FI/ABS(X)
45 A=AN*A1+A0
46 B=4N*B1+B0
47 FI=FI
48 AN=FI+2.
49 AN=FI/ABS(X)
50 A=AN*A1+A0
51 B=4N*B1+B0
52 FI=FI
53 AN=FI+2.
54 AN=FI/ABS(X)
55 A=AN*A1+A0
56 B=4N*B1+B0
57 FI=FI
58 AN=FI+2.
59 AN=FI/ABS(X)
60 A=AN*A1+A0
61 B=4N*B1+B0
62 FI=FI
63 AN=FI+2.
64 AN=FI/ABS(X)
65 A=AN*A1+A0
66 B=4N*B1+B0
67 FI=FI
68 AN=FI+2.
69 AN=FI/ABS(X)
70 A=AN*A1+A0
71 B=4N*B1+B0
72 FI=FI
73 AN=FI+2.
74 AN=FI/ABS(X)
75 A=AN*A1+A0
76 B=4N*B1+B0
77 FI=FI
78 AN=FI+2.
79 AN=FI/ABS(X)
80 A=AN*A1+A0
81 B=4N*B1+B0
82 FI=FI
83 AN=FI+2.
84 AN=FI/ABS(X)
85 A=AN*A1+A0
86 B=4N*B1+B0
87 FI=FI
88 AN=FI+2.
89 AN=FI/ABS(X)
90 A=AN*A1+A0
91 B=4N*B1+B0
92 FI=FI
93 AN=FI+2.
94 AN=FI/ABS(X)
95 A=AN*A1+A0
96 B=4N*B1+B0
97 FI=FI
98 AN=FI+2.
99 AN=FI/ABS(X)
100 A=AN*A1+A0
101 B=4N*B1+B0
102 FI=FI
103 AN=FI+2.
104 AN=FI/ABS(X)
105 A=AN*A1+A0
106 B=4N*B1+B0
107 FI=FI
108 AN=FI+2.
109 AN=FI/ABS(X)
110 A=AN*A1+A0
111 B=4N*B1+B0
112 FI=FI
113 AN=FI+2.
114 AN=FI/ABS(X)
115 A=AN*A1+A0
116 B=4N*B1+B0
117 FI=FI
118 AN=FI+2.
119 AN=FI/ABS(X)
120 A=AN*A1+A0
121 B=4N*B1+B0
122 FI=FI
123 AN=FI+2.
124 AN=FI/ABS(X)
125 A=AN*A1+A0
126 B=4N*B1+B0
127 FI=FI
128 AN=FI+2.
129 AN=FI/ABS(X)
130 A=AN*A1+A0
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132 FI=FI
133 AN=FI+2.
134 AN=FI/ABS(X)
135 A=AN*A1+A0
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137 FI=FI
138 AN=FI+2.
139 AN=FI/ABS(X)
140 A=AN*A1+A0
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142 FI=FI
143 AN=FI+2.
144 AN=FI/ABS(X)
145 A=AN*A1+A0
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148 AN=FI+2.
149 AN=FI/ABS(X)
150 A=AN*A1+A0
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158 AN=FI+2.
159 AN=FI/ABS(X)
160 A=AN*A1+A0
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163 AN=FI+2.
164 AN=FI/ABS(X)
165 A=AN*A1+A0
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169 AN=FI/ABS(X)
170 A=AN*A1+A0
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185 A=AN*A1+A0
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804 AN=FI/ABS(X)
805 A=AN*A1+A0
806 B=4N*B1+B0
807 FI=FI
808 AN=FI+2.
809 AN=FI/ABS(X)
810 A=AN*A1+A0
811 B=4N*B1+B0
812 FI=FI
813 AN=FI+2.
814 AN=FI/ABS(X)
815 A=AN*A1+A0
816 B=4N*B1+B0
817 FI=FI
818 AN=FI+2.
819 AN=FI/ABS(X)
820 A=AN*A1+A0
821 B=4N*B1+B0
822 FI=FI
823 AN=FI+2.
824 AN=FI/ABS(X)
825 A=AN*A1+A0
826 B=4N*B1+B0
827 FI=FI
828 AN=FI+2.
829 AN=FI/ABS(X)
830 A=AN*A1+A0
831 B=4N*B1+B0
832 FI=FI
833 AN=FI+2.
834 AN=FI/ABS(X)
835 A=AN*A1+A0
836 B=4N*B1+B0
837 FI=FI
838 AN=FI+2.
839 AN=FI/ABS(X)
840 A=AN*A1+A0
841 B=4N*B1+B0
842 FI=FI
843 AN=FI+2.
844 AN=FI/ABS(X)
845 A=AN*A1+A0
846 B=4N*B1+B0
847 FI=FI
848 AN=FI+2.
849 AN=FI/ABS(X)
850 A=AN*A1+A0
851 B=4N*B1+B0
852 FI=FI
853 AN=FI+2.
854 AN=FI/ABS(X)
855 A=AN*A1+A0
856 B=4N*B1+B0
857 FI=FI
858 AN=FI+2.
859 AN=FI/ABS(X)
860 A=AN*A1+A0
861 B=4N*B1+B0
862 FI=FI
863 AN=FI+2.
864 AN=FI/ABS(X)
865 A=AN*A1+A0
866 B=4N*B1+B0
867 FI=FI
868 AN=FI+2.
869 AN=FI/ABS(X)
870 A=AN*A1+A0
871 B=4N*B1+B0
872 FI=FI
873 AN=FI+2.
874 AN=FI/ABS(X)
875 A=AN*A1+A0
876 B=4N*B1+B0
877 FI=FI
878 AN=FI+2.
879 AN=FI/ABS(X)
880 A=AN*A1+A0
881 B=4N*B1+B0
882 FI=FI
883 AN=FI+2.
884 AN=FI/ABS(X)
885 A=AN*A1+A0
886 B=4N*B1+B0
887 FI=FI
888 AN=FI+2.
889 AN=FI/ABS(X)
890 A=AN*A1+A0
891 B=4N*B1+B0
892 FI=FI
893 AN=FI+2.
894 AN=FI/ABS(X)
895 A=AN*A1+A0
896 B=4N*B1+B0
897 FI=FI
898 AN=FI+2.
899 AN=FI/ABS(X)
900 A=AN*A1+A0
901 B=4N*B1+B0
902 FI=FI
903 AN=FI+2.
904 AN=FI/ABS(X)
905 A=AN*A1+A0
906 B=4N*B1+B0
907 FI=FI
908 AN=FI+2.
909 AN=FI/ABS(X)
910 A=AN*A1+A0
911 B=4N*B1+B0
912 FI=FI
913 AN=FI+2.
914 AN=FI/ABS(X)
915 A=AN*A1+A0
916 B=4N*B1+B0
917 FI=FI
918 AN=FI+2.
919 AN=FI/ABS(X)
920 A=AN*A1+A0
921 B=4N*B1+B0
922 FI=FI
923 AN=FI+2.
924 AN=FI/ABS(X)
925 A=AN*A1+A0
926 B=4N*B1+B0
927 FI=FI
928 AN=FI+2.
929 AN=FI/ABS(X)
930 A=AN*A1+A0
931 B=4N*B1+B0
932 FI=FI
933 AN=FI+2.
934 AN=FI/ABS(X)
935 A=AN*A1+A0
936 B=4N*B1+B0
937 FI=FI
938 AN=FI+2.
939 AN=FI/ABS(X)
940 A=AN*A1+A0
941 B=4N*B1+B0
942 FI=FI
943 AN=FI+2.
944 AN=FI/ABS(X)
945 A=AN*A1+A0
946 B=4N*B1+B0
947 FI=FI
948 AN=FI+2.
949 AN=FI/ABS(X)
950 A=AN*A1+A0
951 B=4N*B1+B0
952 FI=FI
953 AN=FI+2.
954 AN=FI/ABS(X)
955 A=AN*A1+A0
956 B=4N*B1+B0
957 FI=FI
958 AN=FI+2.
959 AN=FI/ABS(X)
960 A=AN*A1+A0
961 B=4N*B1+B0
962 FI=FI
963 AN=FI+2.
964 AN=FI/ABS(X)
965 A=AN*A1+A0
966 B=4N*B1+B0
967 FI=FI
968 AN=FI+2.
969 AN=FI/ABS(X)
970 A=AN*A1+A0
971 B=4N*B1+B0
972 FI=FI
973 AN=FI+2.
974 AN=FI/ABS(X)
975 A=AN*A1+A0
976 B=4N*B1+B0
977 FI=FI
978 AN=FI+2.
979 AN=FI/ABS(X)
980 A=AN*A1+A0
981 B=4N*B1+B0
982 FI=FI
983 AN=FI+2.
984 AN=FI/ABS(X)
985 A=AN*A1+A0
986 B=4N*B1+B0
987 FI=FI
988 AN=FI+2.
989 AN=FI/ABS(X)
990 A=AN*A1+A0
991 B=4N*B1+B0
992 FI=FI
993 AN=FI+2.
994 AN=FI/ABS(X)
995 A=AN*A1+A0
996 B=4N*B1+B0
997 FI=FI
998 AN=FI+2.
999 AN=FI/ABS(X)
1000 A=AN*A1+A0
1001 B=4N*B1+B0
1002 FI=FI
1003 AN=FI+2.
1004 AN=FI/ABS(X)
1005 A=AN*A1+A0
1006 B=4N*B1+B0
1007 FI=FI
1008 AN=FI+2.
1009 AN=FI/ABS(X)
1010 A=AN*A1+A0
1011 B=4N*B1+B0
1012 FI=FI
1013 AN=FI+2.
1014 AN=FI/ABS(X)
1015 A=AN*A1+A0
1016 B=4N*B1+B0
1017 FI=FI
1018 AN=FI+2.
1019 AN=FI/ABS(X)
1020 A=AN*A1+A0
1021 B=4N*B1+B0
1022 FI=FI
1023 AN=FI+2.
1024 AN=FI/ABS(X)
1025 A=AN*A1+A0
1026 B=4N*B1+B0
1027 FI=FI
1028

SUBROUTINE INUE

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```
A=A1
B=B1
A1=A
B1=B
Q1=Q1
Q1=A/B
IF (ABS((Q1-Q)/(Q1))-1.E-6) 4,4,3
* IF(X) 5,6,6
2 Q1=Q1
6 K=N
7 Q1=X/(FN*X*Q1)
RI(K)=Q1
FN=FN-2.
K=K-1
8 IF(K) 8,8,7
8 FI=ZI
DO 9 I=1,N
FI=FI*RI(I)
9 RI(I)=FI
10 RETURN
END
```

INUE 560
INUE 570
INUE 580
INUE 590
INUE 600
INUE 610
INUE 620
INUE 630
INUE 640
INUE 650
INUE 660
INUE 670
INUE 680
INUE 690
INUE 700
INUE 710
INUE 720
INUE 730
INUE 740
INUE 750
INUE 760

REGISTER ALLOCATION

3 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 66
1 REGISTERS ASSIGNED OVER THE LOOP BEGINNING AT LINE 72

FUNCTION RANF CDC 6600 FTM V3.0-P213 OPT=2 03/26/72 01.12.49. PAGE 1

```

5      C      FUNCTION RANF (NS,MJD,C)
      C      DIMENSION NS(2),NC(2)
      C      COMMON /RAN/ N1, N2, MP, T1, T2
      C      DATA M1, M2/24,73-, 155501/
      C      MODE= 10 CONTINUE, OTHERWISE RESTART WITH
      C      *INTEGER NUMBER NS(1)*2**10*NS(2)
      C      IF (MODE) 10, 100, 10
      C      10 N1=NS(1)
      C      N2=NS(2)
      C      T1=2.**(-10)
      C      T2=2.**(-30)
      C      MP=2**18
      C      1 DO 20 I=1,2
      C      GO TO (1,2),I
      C      100 K=12*N2
      C      GO TO 19
      C      12 K=M1*N2+12*N1+KJ
      C      19 KD=K/MP
      C      2 NC(I)=K-KD*MP
      C      N1=NC(2)
      C      N2=NC(1)
      C      XN1=N1
      C      XN2=N2
      C      RANF=XN1*T1+XN2*T2
      C      RETURN
      C      END

```

SYMBOLIC REFERENCE MAP

ENTRY POINTS	DEF LINE	REFERENCES	RELOCATION	SN	TYPE	REFS
2 RANF	1	25				
VARIABLES						
76 I					INTEGER	REFS
77 K					INTEGER	REFS
100 K0					INTEGER	REFS
0 MODE					INTEGER	REFS
2 MP					INTEGER	REFS
72 M1					INTEGER	REFS
73 M2					INTEGER	REFS
103 NC					INTEGER	REFS
0 NS					INTEGER	REFS
0 N1					INTEGER	REFS
1 M2					INTEGER	REFS
3 T1					REAL	REFS
4 T2					REAL	REFS
101 XN1					REAL	REFS
102 XN2					REAL	REFS
STATEMENT LABELS						
0 L0					INACTIVE	REFS
27 103					INACTIVE	REFS
41 11					INACTIVE	REFS
44 12					INACTIVE	REFS
51 193					INACTIVE	REFS
0 203					INACTIVE	REFS

STATEMENT LABELS	DEF LINE	REFERENCES
0 L0		2*7
27 103		13
41 11		15
44 12		17
51 193		19
0 203		19

LOOPS	LABEL	INDEX	FROM-TO	LENGTH	PROPERTIES
33	200	* 1	13 19	258	Opt
COMMON BLOCKS					
RN					
MEMBERS					
1 N1					(1)
3 T1					(1)

STATISTICS	PROGRAM LENGTH	COMMON LENGTH	1129	58	74	5
1 N2						(1)
4 T2						(1)

SUBROUTINE GAUSS

CDC 6600 FTN V3.0-P213 OPT=2 03/26/72 -1.12.49.

PAGE

1

```

SUBROUTINE GAUSS(JS,SD,XM,X)
  DIMENSION NST(2)
  COMMON /RN/ N1,N2,MC,I1,I2
  COMMON /GN/ THOPT,J,XR(2)
  IF (J) 10, 1, 20
10 J=2
  THOPT=8.,ATAN(1.)
  NST(1)=244734
  NST(2)=158551
  NST(1)=102943
  NST(2)=105617
  XR(1)=RANF(NST,1)
  GO TO 35
20 GO TO (30,40), J
30 J=2
  XR(1)=RANF(NST,J)
  XR(2)=RANF(NST,J)
  X1=SQRT(ABS(-2.*ALOG(XR(1))))
  XR(2)=THOPT*XR(2)
  XR(1)=X1*SIN(XR(2))
  XR(2)=X1*COS(XR(2))
  X=XR(1)*SD+XM
  RETURN
4 J=1
  X=XR(2)*SD+XM
  RETURN
  END
```

SUBROUTINE GAUSS

SYMBOLIC REFERENCE MAP

CUC 6600 FTN V3.0-P213 OPT=2 03/26/72 C1.12.49.

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ENTRY POINTS	DEF LINE	REFERENCES	20
2 GAUSS	1	23	
VARIABLES	SN	TYPE	RELOCATION
1 J	INTEGER	GN	
0 JS	INTEGER	*UNUSED	
2 MC	INTEGER	F.P.	
400 NST	INTEGER	RN	
		ARRAY	
0 N1	INTEGER	RN	
1 N2	INTEGER	RN	
0 S0	REAL	F.P.	
0 TMOPI	REAL	GN	
3 T1	REAL	RN	
4 T2	REAL	RN	
0 X	REAL	F.P.	
0 XM	REAL	F.P.	
2 XR	REAL	GN	
		ARRAY	
77 X1	REAL		

EXTERNALS

ALOG	REAL	TYPE	ARGS	DEF LINE	REFERENCES
ATAN	REAL	1 LIBRARY	1	18	
COS	REAL	1 LIBRARY	1	7	
RANF	REAL	1 LIBRARY	2	21	
SIN	REAL	1 LIBRARY	1	12	
SQRT	REAL	1 LIBRARY	1	2	

INLINE FUNCTIONS	TYPE	ARGS	DEF LINE	REFERENCES
ABS	REAL	1	INTRIN	18

STATEMENT LABELS

STATEMENT LABELS	DEF LINE	REFERENCES
0 10	INACTIVE	6
16 20		14
24 30		15
30 35		17
53 40		24

COMMON BLOCKS

COMMON BLOCKS	LENGTH	MEMBERS - BIAS NAME(LENGTH)
RN	5	1 N1 (1)
GN	4	3 T1 (1)
		1 T2 (1)
		1 J (1)

STATISTICS

STATISTICS	PROGRAM LENGTH	COMMON LENGTH
	1728	66
	118	9

Additional Appendix F.

Some Unpublished Conference Papers
Referenced by this Report

The papers in this Appendix are referenced frequently in this report. Because of their relative inaccessibility to the readers of the report they are reproduced for reference here.

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REALIZATION OF NON-LINEAR FILTERS*

R. S. Bucy**

Abstract

The concept of numerically exact non-linear filter synthesis is defined and shown to be of importance. Confidence intervals for error performance are described, which allow one to make meaningful statistical inferences about filter performance. Further, some recent synthesis results for non-linear filters are discussed.

1. INTRODUCTION

In the past eleven years, a rather amazingly complete theory of non-linear filtering has emerged, see [1], [2], [3], and [4]. In fact, the representation theorem, see [4], has provided a solution to a general class of non-linear filtering problems, albeit the solution depends on the evaluation of a function space integral.

Unfortunately, the applications of this theory have been almost non-existent. The reason for this state of affairs is paradoxical; the success and wide-spread use of the Kalman-Bucy linear theory, see [4] and [5]. The situation is reminiscent of the popularity of the frequency domain techniques in the mid-50's, which delayed the state space viewpoint for some years.

In this paper, I would like to review some of the efforts at synthesis of optimal filters for some specific problems, and at the same time point out some slightly paradoxical things I have learned over the past three years. In particular, the meaning of Monte Carlo results will be discussed in some detail, as it is now commonplace to design systems on the basis of Monte Carlo runs,

too often only 25 to 100 Monte Carlo runs are made. Of course, as we shall show, such a small number of runs produces unbelievably large uncertainties. Finally, we will discuss in detail a parameter estimation problem considered by Licht in [6] and the phase lock study in [7].

2. NUMERICALLY EXACT REALIZATIONS

The basic problem of machine realization of optimal non-linear filters consists in the iteration of a non-linear convolution equation of the form

$$J_n(x) = \int_{-\infty}^{\infty} T_n(x, \xi) J_{n-1}(\xi) d\xi \quad (2.1)$$

with $J_n(x)$ the conditional density of the signal process at time n , given the observations up to $n-1$ and T_n a non-linear function of the last piece of data. Of course, (2.1) is nothing but a discrete form of Bayes' rule which can be written as (2.1), when the signal process is Markovian and the plant noise as well as the observation noise are independent white noise sequences. For details on how $J_n(x)$ can be represented as a set of $(2M+1)^{d^{***}}$ weights on a moving grid, see [1] and [8]. The important point is that the point

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*** d is the state dimension.

mass representation is numerically exact, that is the sup of the difference between the true conditional probability of a compact set and the approximating probability converges to zero as the number of point masses gets large.

To illustrate the importance of numerical exactness, consider the following problem:

$$\begin{aligned} x_n &= \alpha x_{n-1} + u_{n-1}, \quad x_0 = c \\ z_n &= x_n^3 + v_n \end{aligned} \quad (2.2)$$

with u_n and v_n independent zero mean Gaussian white noise sequences and c a mutually independent zero mean Gaussian white noise sequence. This problem was considered in [11] and [10]. The behavior of $\hat{x}_{n/n}(M) = E x_n | z_1, \dots, z_n$ as computed as function of M , the number of weights in the approximation to $J_n(x)$ is quite remarkable. As long as none of the observations are large in comparison to the observation noise standard deviation, an M of 48 produces $\hat{x}_{n/n}(M) \approx \hat{x}_{n/n}(2M)$ to four places. However, if z_n is large, then $\hat{x}_{n/n}(256) = z_n^{1/3}$. What happens is that in the case of z_n large, the filter density approximates a delta function and the weight approximation to the filter density must contain a large number of weights to be accurate. The phenomenon is quite interesting as it shows that bandwidth is irrelevant for non-linear filters. Because we used a numerically exact realization scheme, we found this phenomenon, which, by the way is easy to explain after it was observed by looking at the error of the estimate $z_n^{1/3}$.

Without a numerically exact scheme, there is no way of knowing when the approximation to the optimal filter is adequate. We note in passing that the sub-optimal schemes, such as [12] are not numerically exact, as T_n is replaced by a truncated series approximation.

3. MONTE CARLO ERROR EVALUATION

Suppose that a random variable x is estimated as \hat{x} and that x_i and \hat{x}_i denote the i independent

trial and the statistic S_n^2 is computed as

$$S_n^2 = \frac{1}{n} \sum_{i=1}^n \tilde{x}_i^2$$

with $\tilde{x}_i = x_i - \hat{x}_i$, then

$$\frac{N S_n^2 - N \mu_2}{\sqrt{N} \sqrt{\mu_4 - \mu_2^2}} \quad \text{with } \mu_\alpha = E \tilde{x}^\alpha \quad (3.1)$$

is asymptotically normal of mean zero and variance 1 in view of the central limit theorem.

In other words, S_n^2 is normal mean μ_2 and variance $\frac{(\mu_4 - \mu_2^2)}{n}$.

For Monte Carlo error analysis then S_n^2 is the estimate of the mean square error, and with probability .9972

$$-3 \sqrt{\frac{\mu_4 - \mu_2^2}{n}} \leq S_n^2 - \mu_2 \leq 3 \sqrt{\frac{\mu_4 - \mu_2^2}{n}} \quad (3.2)$$

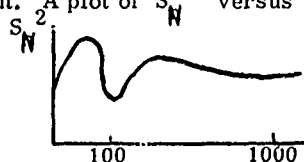
if $\mu_4 \approx 3\mu_2^2$ as it is when \tilde{x} is Gaussian, then with probability .9972

$$\frac{S_n^2}{1 + 3\sqrt{\frac{2}{n}}} \leq \mu_2 \leq \frac{S_n^2}{1 - 3\sqrt{\frac{2}{n}}} \quad (3.3)$$

In Figure I, we have plotted $3\sqrt{\frac{2}{n}}$ versus N . Note how slowly $3\sqrt{\frac{2}{n}}$ goes to zero. The net effect is that in order to achieve a 3σ confidence interval of overall length .2 of S_n^2 , 2,000 Monte Carlo's are needed; under the assumption $\mu_4 \approx 3\mu_2^2$. Of course, if $\mu_4 < 3\mu_2^2$, the situation improves, but if $\mu_4 > 3\mu_2^2$, the situation is worse.

In view of these results, the Monte Carlo analysis of systems which is often made with 16 to 100 runs, seems curious at best. Of course, another advantage of synthesis methods, which are numerically exact and carry the distribution, is that μ_4 can be at least approximately determined, so that with (3.3) the confidence of an estimate can be more exactly determined. We used 2,000 Monte Carlo runs to evaluate the performance of the phase-lock loop in [7] while 500 Monte Carlo runs were used in [8], as we

proved order of magnitude mean square error betterment. A plot of S_N^2 versus N can look like



and if such a graph is made for a particular problem, it will quickly convince one that no inference should be made on μ_2 without a bare minimum of about 500 Monte Carlo runs.

In order to illustrate that the problems raised above are real, we consider the following problem considered first by Licht in [6]

$$\frac{dx}{dt} = -x, \quad x(0) = c \quad (3.4)$$

$$dz = (x + x^3)dt + dv$$

with c Gaussian mean one variance 1 and the dv process white noise of spectral density R . Licht used the Representation Theorem and obtained the experimental results shown in Figures II, III and IV, which he disclosed to me in [13], after I had seen the results of [6] and suggested he simulate the sensor orbit filter. The sensor orbit filter, see [4] and [9], is the wide sense Kalman-Bucy filter for the system

$$\begin{aligned} \frac{dy_1}{dt} &= -y_1 & \text{and} & \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} \in N \left(\begin{pmatrix} 1 \\ 4 \end{pmatrix}, \begin{pmatrix} 1 & 6 \\ 6 & 60 \end{pmatrix} \right)^* \\ \frac{dy_2}{dt} &= -3y_2 \\ dz &= (y_1 + y_2)dt + dv \end{aligned} \quad (3.5)$$

where $y_1 = x$, $y_2 = x^3$. Note that (3.5) is just (3.4) in new coordination. The true error performance of the wide-sense filter for (3.5) is easily evaluated by solving a Riccati equation and is shown on the figures. Notice first if we assume that $\mu_4 \geq 3\mu_2^2$ that the three plots are inconclusive about the relative error performance of

the optimal versus the extended Kalman-Bucy filter as the confidence intervals are too large. In fact, none of the experimental results are precise enough to draw any conclusions**. However, the true sensor orbit performance is significantly higher than the experimental extended Kalman-Bucy filter error, on the basis of Figure II, only. The conclusion that we may draw then is the following, the extended K-B filter performance depends on the coordination of the signal process, not too surprising. However, even more significant is the following: The performance of the extended K. B. filter may be worse in sensor orbit coordinates. (N. B. the wide-sense sensor orbit filter coincides with the extended K. B. filter for (3.5)).

This result surprised me, as intuitively I expected the extended K. B. filter to perform better in sensor orbit coordinates. I feel it necessary to remark that Licht's thesis is an excellent piece of work; its only defect consists in drawing conclusions from too small a number of Monte Carlo runs. This defect is shared by many papers, for example [12], where 50 Monte Carlo's are used.

3. PHASE LOCK LOOP DESIGN

A problem which has an interesting history as well as major technological importance is that of phase detection. An idealized model of which is the following

$$\begin{aligned} d\phi &= d\beta; \quad \phi_0 = c \\ dz_1 &= \cos \phi dt + dv_1 \\ dz_2 &= \sin \phi dt + dv_2 \end{aligned} \quad (4.1)$$

where $d\beta$, dv_1 , dv_2 are mutually independent white noise processes of spectral densities q , $2r$, and $2r$, while c is a Gaussian random variable of zero mean and variance A . A more realistic model, ϕ the output of two integrators is being

* $y \in N(\mu, P)$ denotes that y is normally distributed mean μ covariance P .

** The confidence intervals are too large.

studied by Hecht in [10], however, I will confine myself to a discussion of the model (4.1). The interested reader should consult [7] for the detailed analysis of the optimal filter for (4.1).

It can easily be shown that the steady-state extended K-B filter for the model (4.1) is given by

$$d\hat{\phi} = \sqrt{\frac{q}{2r}} (dz_2 \cos \hat{\phi} + dz_2 \sin \hat{\phi}) \quad (4.2)$$

with $\hat{\phi}(0) = 0$.

Now $\hat{\phi}$ given by (4.2) is the output of the classical 1 dimensional phase-lock loop and for small $\phi - \hat{\phi}$ the error is $R = \sqrt{2rq}$.

Since for an important class of problems only the relative phase is of interest, in [7] we considered the loss function $L(\phi - \phi^*) = 2(1 - \cos(\phi - \phi^*))$ which is quadratic for small $\phi - \phi^*$, but cyclic otherwise, in other words, ϕ^* was chosen so that

$$\begin{aligned} E L(\phi_n - \phi_n^*) | z_i(j) \quad i=1, 2; j=1, \dots, n-1) \\ \leq E L(\phi_n - \psi) | z_i(j) \quad i=1, 2; j=1, \dots, n-1) \end{aligned} \quad (4.3)$$

for the discrete version of the problem. It is easily seen that ϕ_n^* is given by

$$\phi_n^* = \tan^{-1} \frac{\hat{S}_n}{\hat{C}_n} \quad (4.4)$$

with $\hat{S}_n = E \sin \phi_n | z_i(j) \quad i=1, 2; j=1, \dots, n-1)$
 $\hat{C}_n = E \cos \phi_n | z_i(j) \quad i=1, 2; j=1, \dots, n-1).$

Let us agree to call ϕ_n^* the cyclic non-linear estimator. It is easily seen that

$$\begin{aligned} E L(\phi_n - \phi_n^*) | z_i(j) \quad i=1, 2; j=1, \dots, n-1) \\ = 2(1 - \sqrt{\hat{C}_n^2 + \hat{S}_n^2}) \end{aligned} \quad (4.5)$$

In Figure V, the experimental results are given together with the theoretical error curve derived by Viterbi in [14]. The 3σ confidence interval extends .4 dB above and below the non-linear error performance curve, so that we may conclude that the optimal filter achieves a mean square error improvement of between .6 dB and 1.4 dB.

However, since the lower curve represents a lower bound, the maximum achievable error improvement would be 2 dB. The optimal filter achieves 1/3 to 2/3 of the error performance difference of the filter which senses ϕ directly versus the classical phase-lock loop. The maximum achievable error improvement for the more realistic two-dimensional phase-lock loop (i.e., the phase is doubly integrated white noise) is about 6 db, so that one might hope for a 3 db error betterment for the optimal filter for this problem, and this would certainly affect the way on how phase-lock loops are built in the future. Unfortunately, the current economic doldrums have adversely affected NASA's ability to support such a study.

4. CONCLUSIONS

In the past three years, non-linear optimal filters have been built, using digital computers, and have been evaluated, using Monte Carlo results. The purpose of this paper has been to indicate the amount of Monte Carlo runs necessary to make valid statistical inferences as to filter performance, as well as reviewing some of the results of recent studies of non-linear filters. I hope this paper will have the effect of stimulating meaningful work in the area of non-linear filtering.

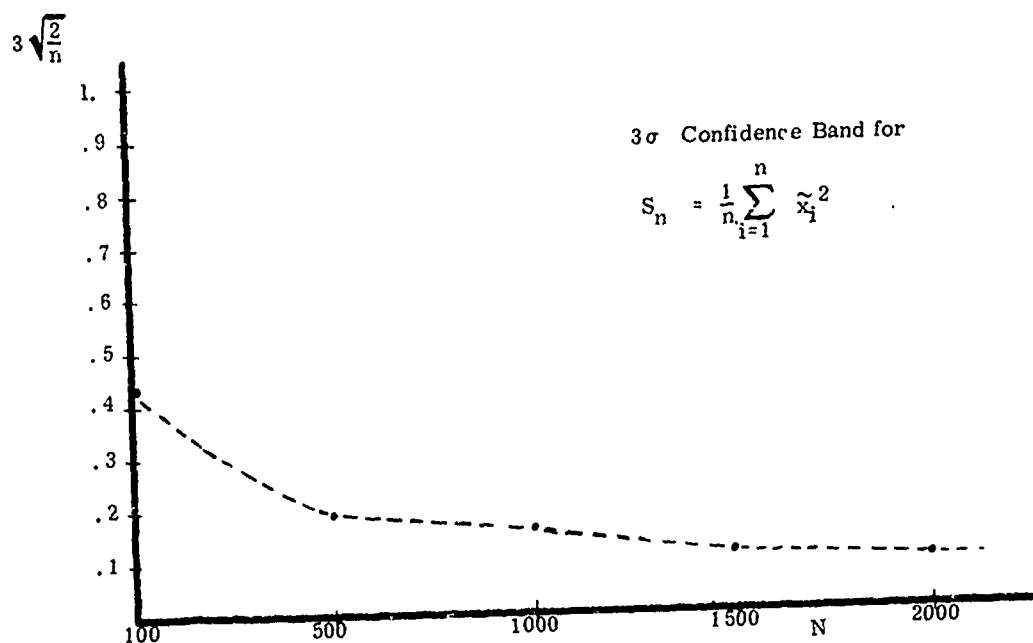


FIGURE I

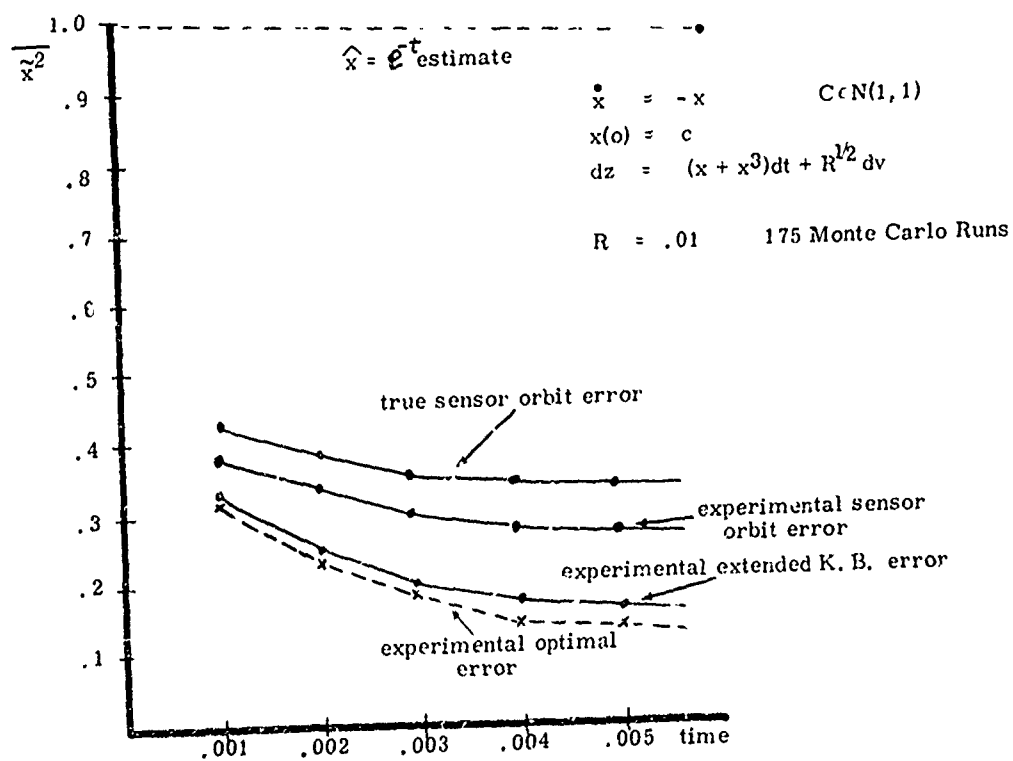


FIGURE II

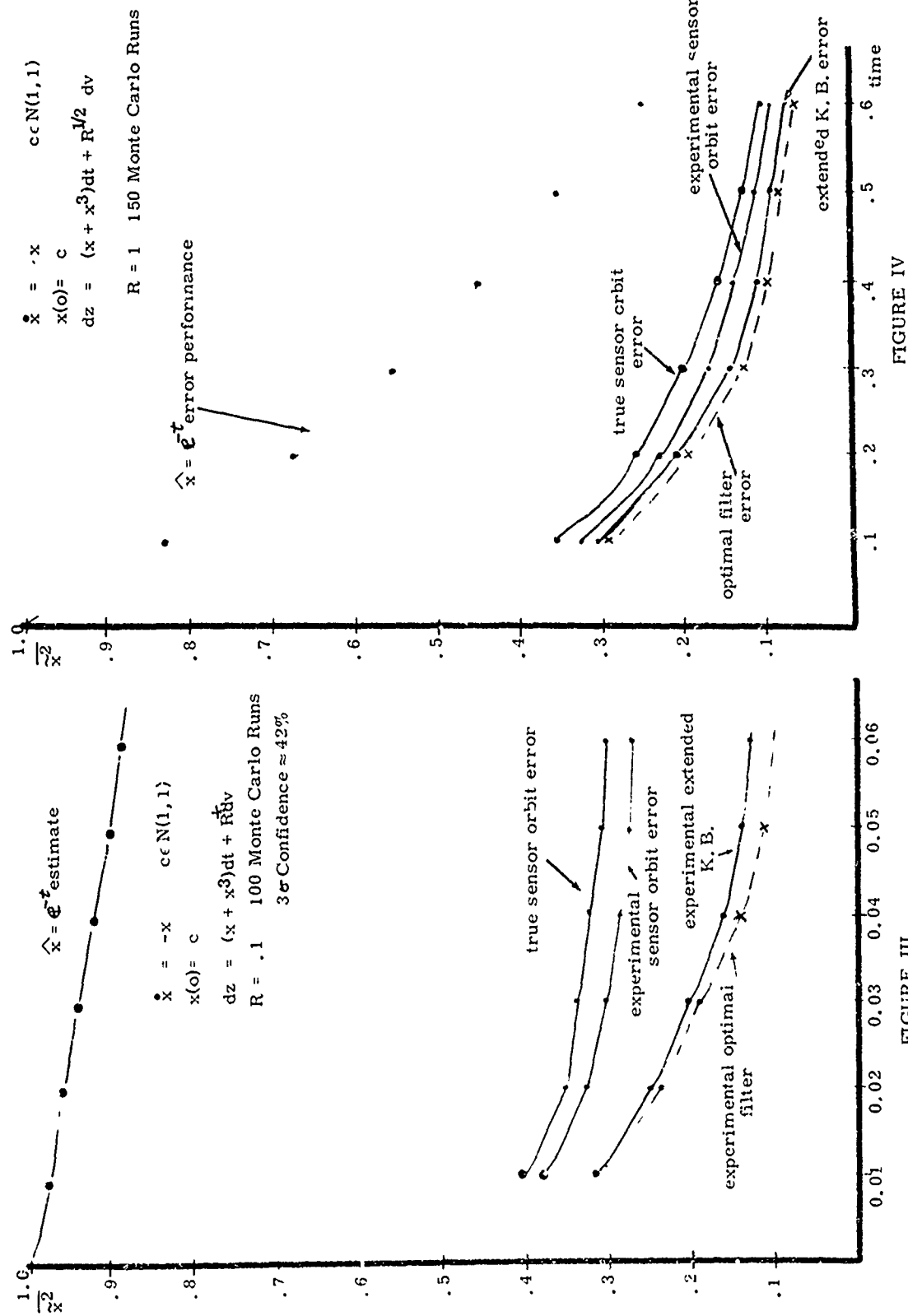


FIGURE III

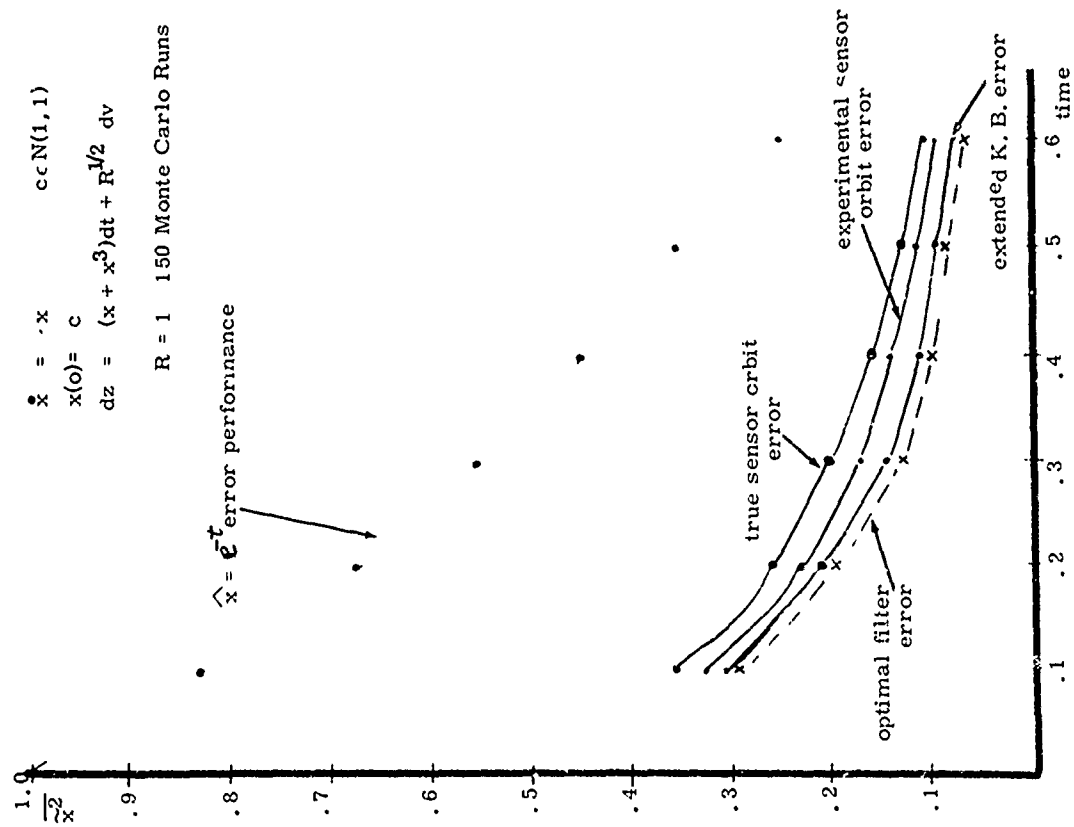


FIGURE IV

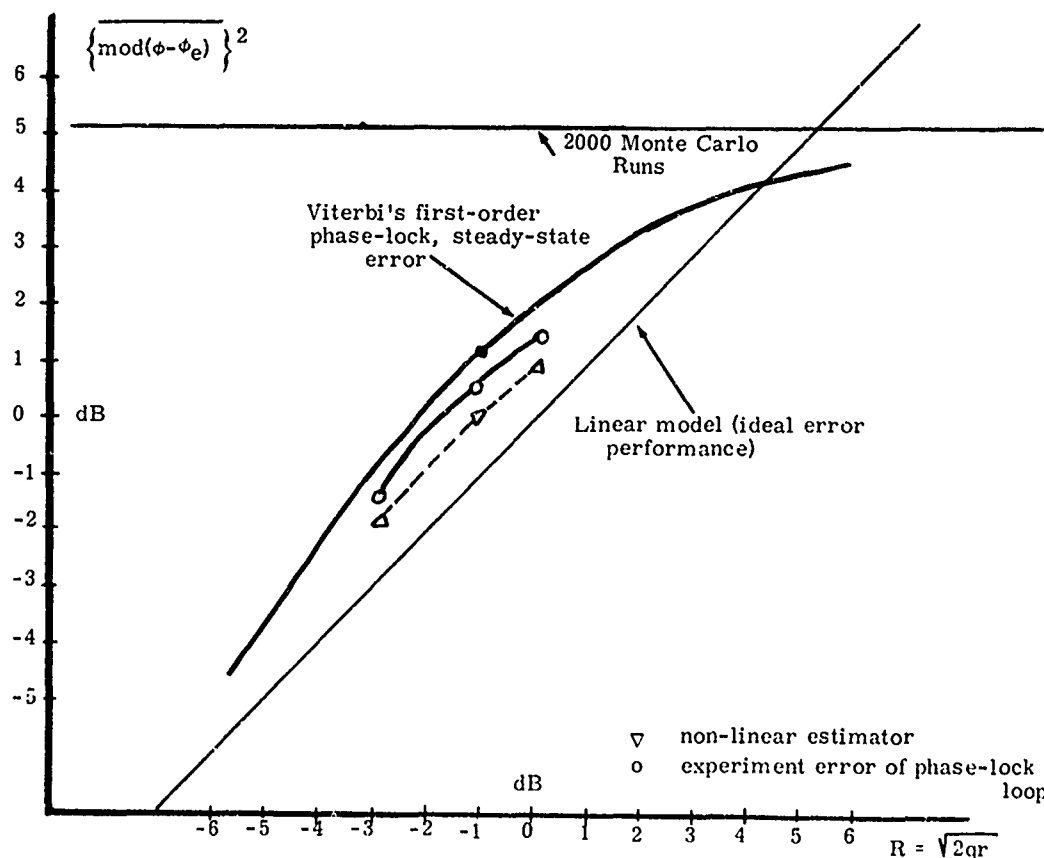


FIGURE V

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HYBRID COMPUTER SYNTHESIS of OPTIMAL DISCRETE NONLINEAR FILTERS*

R. S. Bucy, M. J. Merritt and D. S. Miller**

Abstract

The total number of digital operations required to generate an estimate using a discrete optimal nonlinear filter is extremely large. This paper delineates the computer task and introduces an algorithm for mechanizing an optimal 1-step predictor using a hybrid computer to reduce the computation time. The advantages and limitations of the hybrid mechanization are described. Hybrid computer predictions (as simulated by a continuous system simulation language) are compared to those generated by an all digital mechanization. Timing estimates, accuracy estimates and computer equipment requirements are given. Extensions of the algorithm to multidimensional systems and observations are discussed. Use of a parallel digital multiprocessor to perform the algorithm is demonstrated. The use of interactive computer graphics for debugging and control of the algorithm is discussed.

1. INTRODUCTION

It has been known for a long time that the optimal, discrete time, nonlinear estimator for systems whose plant dynamics, sensor characteristics and signal statistics are known is just the algorithm obtained using Bayes' Rule to sequentially update the conditional probability density based on the latest data. However, until recently researchers have avoided this numerically exact procedure, preferring instead to develop suboptimal Taylor Series estimators, as for example the extended Kalman-Bucy filter and the minimum least-squares error second order filter. The reason for this is clear. A true optimal filter is a computational juggernaut. In the past few years the availability of 3rd generation high speed digital processors has led to several papers by Bucy (3), Bucy and Senne (4), (5), and Bucy, Senne and

Geesey (6) which describe digital computer mechanizations for the required iteration of the conditional probability density function. Both one and two state variable nonlinear problems with Gaussian statistics have appeared in the literature. It is quite clear that an extremely simple 4 state variable filtering problem with a real time data rate of 1 sample per second pushes contemporary high speed sequential processors to the limits of their capabilities.

The hard core of the problem in computing the $(n+1)$ st estimate is the need to re-evaluate the d -dimensional conditional probability density function $J_{n+1}(y)$ for each piece of data. This requires the evaluation of a d -fold convolution integral:

$$J_{n+1}(y) = \int_{R^d} T_n(y, \xi) \cdot J_n(\xi) d\xi \quad (1)$$

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If the statistics are Gaussian, $T_n(y, \xi)$ is an exponential with vector-matrix argument. $J_n(\xi)$ is the previous conditional density function evolved to time t_n . If the J_n 's are M -ly discretized in all directions then $J_n(\xi)$ and $J_{n+1}(y)$ are each defined on a grids containing M^d points. The computation requires M^d evaluations of the integrand for each of the M^d points in the domain of $J_{n+1}(y)$. Thus one iteration of the algorithm requires $(M^d)^2 = M^{2d}$ exponentiations each of which entails the computation of a multitermed quadratic form. The integration and statistical computations which follow the convolution consume considerably less computer time proportionally. Even for low dimensionality filters: 2 state variable (2-D) plant with linear additive plant and observation noise sequences, Bucy and Senne, (4), introduced rotated coordinates and ellipsoidal mode tracking to achieve a real time data rate of one prediction per two seconds. Suppose the conditional density at time t_n : $J_n(\xi)$, and the corresponding element of the discrete data sequence z_n , are available. The computation may be speeded up by generating the exponential portions of the integrand both simultaneously and continuously. This is easily done on an analog computer. However, the analog computer is notably poor at storage of time varying functions of several variables. Analog computers are not well adapted to arithmetic computations, the solution of algebraic equations, etc. Hybrid computers were developed to overcome these problems by combining the digital computer's arithmetic, storage and control functions with the analog computer's ability to solve large numbers of simultaneous differential equations rapidly. As might be expected, hybrid computation introduces new problems. Some of the most important of these are:

- (1) How to partition equations between analog and digital computers,
- (2) Development of analog representations of filter variables,
- (3) Setting of initial conditions,
- (4) Effects of limited bandwidth of analog computer components,
- (5) Effects of digital to analog and analog to digital conversion times,
- (6) Limitations imposed by maximum digital computer I/O data rates,
- (7) Delays caused by analog computer mode switching and reset times,
- (8) The effect of limited analog computer dynamic range.

These problems are considered in later sections of the paper, and estimates of performance of a high speed hybrid computer (containing a contemporary analog computer and a 3rd generation high speed digital processor) are given.*

The sections below will describe:

- (1) The computer task in optimal filter synthesis,
- (2) Review the digital computer algorithm proposed by Bucy (Ref. 3) for a one state variable (1-D) problem with linear plant and cubed sensor,
- (3) Describe the hybrid algorithm, illustrating its application to the same cubed sensor problem. The basic theory and flow charts are presented.
- (4) Some of the details which lead to a discussion of the hybrid algorithm's speed

*In mid-1971 a contemporary analog computer might be an Electronic Associates EAI 680, Applied Dynamics AD-4 or Comcor Ci-5000. The high speed digital processor might be a Control Data CDC 6600 or IBM 360/85.

and accuracy are described.

- (5) The latest results obtained for the cubed sensor problem (simulating the hybrid algorithm with a digital continuous system simulation language) are presented.
- (6) We will briefly discuss some of the interesting extensions of the above: 2-D and multi-D problems, the effects of nonlinear plant on grid spacing and grid floating; and the use of a parallel digital multiprocessor such as the ILIAC IV to implement the hybrid algorithm.
- (7) Preliminary timing estimates for 1-D and 2-D filters for selected hardware configurations are given and compared against similar estimates for the all digital algorithm.
- (8) The use of interactive computer graphics for display and debugging of optimal filters and their density functions is described.
- (9) Finally, we summarize the advantages and disadvantages of the hybrid algorithm as it compares to the purely digital optimal filter realization and conclude with a description of work in progress.

2. THE COMPUTER TASK IN OPTIMAL FILTER SYNTHESIS

Optimal discrete nonlinear filters are mechanized as an iterative process wherein the same basic procedure is recomputed for each new piece of data. During each iteration, a new conditional density function and a new conditional mean are evolved. Many substeps and numerous calculations are involved. For convenience and comparative purposes, only the one-step predictor is considered here. This section reviews the all digital algorithm of Bucy (3) and its application to a one-dimensional (1-D) cubed sensor problem.

Subsequent sections describe the hybrid mechanization.

In the all digital algorithm, for computational purposes, the conditional probability density function is represented as a set of point masses, i.e., a set of M dirac delta functions appropriately positioned and weighted, see Figure (1). In order to obtain the weighting to be assigned to each point in the new grid space, a convolution must be performed over every point carrying positive mass in the old grid space. This convolution is, by far, the most time consuming portion of the calculation and, in essence, is "THE COMPUTER TASK" in optimal filter synthesis. If the running variables of the conditional density function J_n are finely discretized over a large domain, then the computation can become burdensome. If the grid is allowed to "float" at the beginning of each iteration, it need not be of such high resolution nor broad span as would otherwise be required. This leads to a considerable reduction in computer workload. In (3), the grid for J_{n+1} is centered at the mean of J_n , \hat{x}_n , and adjusted to span $6\hat{\sigma}_n$ or $8\hat{\sigma}_n$ to either side, where $\hat{\sigma}_n$ is the standard deviation of J_n .

2.1 THE ONE DIMENSIONAL CUBED SENSOR PROBLEM*

The model:

$$\begin{aligned} x_n &= ax_{n-1} + u_{n-1} \\ x_0 &= c \\ z_n &= x_n^3 + v_n \end{aligned} \quad (2.1)$$

The signal process $\{x_n\}_{n=1, \dots, m, \dots}$ is a discrete time, indexed set of scalar valued random variables with $a > 0$.

The stochastic process $\{u_n\}_{n=1, \dots, m, \dots}$ is a set of independent, normal, identically distributed scalar valued random variables, with variance, q . Hence the conditional density of the $(n+1)$ st state, y , given the n^{th} state is:

$$f_{y|x_n=\xi}(\xi) = \frac{1}{\sqrt{2\pi q}} e^{-\frac{1}{2q} [y-a\xi]^2} \quad (2.2)$$

*A more detailed development of this problem is given by Bucy in (3).

The random variable c is a scalar independent of the u_n process and having a density

$$f_c(\xi) = \frac{1}{\sqrt{2\pi k}} e^{-\frac{1}{2k}(\xi)^2} \quad (2.3)$$

The observation process $\{z_n\}_{n=1, \dots, m, \dots}$ is a scalar valued data sequence. The stochastic process $\{v_n\}_{n=1, \dots, m, \dots}$ is a set of independent, normal, identically distributed scalar random variables, with variance, r , and independent of the u_n process and the random variable c . Hence the conditional density of the n^{th} observation given the n^{th} state is:

$$f_{z_n|x_n} = \xi(\xi) = \frac{1}{\sqrt{2\pi r}} e^{-\frac{1}{2r}(z_n - \xi)^2} \quad (2.4)$$

Based on this description of the signal and observation processes, a sequential application of Bayes' Rule leads to the following update relationship for the probability density function representing x_{n+1} at time t_{n+1} conditioned on data z_n, \dots, z_0 which we abbreviate as $J_{n+1}(y)$.

$$J_{n+1}(y) \stackrel{\circ}{=} \int_{-\infty}^{\infty} e^{-\frac{1}{2q}(y - a\xi)^2} e^{-\frac{1}{2r}(z_n - \xi)^2} \times J_n(\xi) d\xi \quad (2.5)$$

where $\stackrel{\circ}{=}$ means equality to a normalization constant dependent on z_{n-1}, \dots, z_0 .

There are essentially two problems associated with implementing an optimal nonlinear estimator:

- (1) It is necessary to represent the non-parametric conditional density in some practical storage media - the representation problem.
- (2) Evaluation of the integral, equation (2.5), - the convolution problem.

2.2 REPRESENTATION

Bucy (3) describes the probability density function as M ordered pairs $\{J_n[\xi_n(i)], \xi_n(i)\}$, $i = 1, \dots, M$ with: $\xi_n(i)$ a map from the set $\{1, 2, \dots, M\}$ to the reals; $J_n[\xi_n(i)]$ is the corresponding value of J_n

a $\xi_n(i)$. Considering the components of J_n as nonnegative masses, $\xi_n(i)$ is the point in R^1 which carries the i^{th} mass $J_n[\xi_n(i)]$.

The continuous probability density function $J_n(\cdot)$, is thus discretely representable by M delta functions of weight $J_n[\xi_n(i)]$ located at $\xi_n(i)$, $i=1, \dots, M$.

Now define

$$\begin{aligned} y_i &\triangleq \xi_{n+1}(i) \\ \xi_j &\triangleq \xi_n(j) \end{aligned} \quad (2.6)$$

and (2.5) can be rewritten as a discrete density function:

$$J_{n+1}(y_i) \stackrel{\circ}{=} \sum_{j=1}^M e^{-\frac{1}{2q}(y_i - a\xi_j)^2} e^{-\frac{1}{2r}(z_n - \xi_j)^2} J_n(\xi_j) \quad (2.7)$$

with $i = 1, \dots, M$

where $\sum_{i=1}^M J_{n+1}(y_i) \triangleq 1$.

For an initial condition from (2.3),

$$J_0(\xi_j) = e^{-\frac{1}{2k} \xi_j^2} \quad j=1, \dots, m. \quad (2.8)$$

The gridding $\xi_n(j)$ is centered at mean \hat{x}_{n-1} and spans $N \cdot \hat{\sigma}_{n-1}$ standard deviations to either side where N is determined empirically. Hence, we have

$$\begin{aligned} \xi_{n+1}(i) &= \hat{x}_n - N\hat{\sigma}_n + \frac{N\hat{\sigma}_n}{(M-1)/2} \cdot (i-1) \\ &\text{with } 1 \leq i \leq M \end{aligned} \quad (2.9)$$

$$\begin{aligned} \xi_1(i) &= \xi_0(i) \text{ with (from (2.3)) } \hat{x}_0 = 0 \text{ and} \\ \hat{\sigma}_0 &= \sqrt{k}. \end{aligned}$$

With the representation (2.6)-(2.9), the convolution (2.5) is solved for $J_{n+1}(y)$. After normalization, its mean and variance are computed.

2.3 CONVOLUTION

Given the representation (2.6)-(2.9), (2.5) is solved for $J_{n+1}^U(y)$ in M sequential integrations each

consisting of M integrand evaluations, where

$J_{n+1}^U(y)$ is $J_{n+1}(y)$ prior to normalization.

2.4 ESTIMATE COMPUTATION

After normalization to obtain $J_{n+1}(y)$, computation of its mean, \hat{x}_{n+1} , and variance $\hat{\sigma}_{n+1}^2$ are straightforward:

$$\hat{x}_{n+1} = \sum_{i=1}^M y_i J_{n+1}(y_i)$$

$$\hat{\sigma}_{n+1}^2 = \sum_{i=1}^M [y_i^2 J_{n+1}(y_i)] - (\hat{x}_{n+1})^2$$

An overall flow chart of the digital algorithm for this problem is shown in figure (2). The integrand evaluation and computation of estimates are shown in figures (3) and (4). An expanded flow chart, figure (5), shows the integrand evaluation for a 2-D optimal 1-step predictor and is included as a further illustration that convolution, and in particular, integrand evaluation is the problem in discrete optimal filtering.

3. THE HYBRID ALGORITHM

The preceding discussion has pinpointed the major obstacle to high speed real time filtering. This obstacle is the need to calculate the conditional density functions and their corresponding argument values M^{2d} times for every piece of new data. For example, in the cubed sensor problem the integrand must be computed M^2 times during the M evaluation of the discrete convolution

$$J_{n+1}(y_I) \cong \sum_{j=1}^M e^{-\frac{1}{2q} [y_I - a\xi_j]^2} e^{-\frac{1}{2r} [z_n - \xi_j^3]^2} \chi$$

$$J_n(\xi_j) \quad (3.1a)$$

which is a discrete approximation to the continuous convolution

$$J_{n+1}(y) \cong \int_{-\infty}^{\infty} e^{-\frac{1}{2q} (y - a\xi)^2} e^{-\frac{1}{2r} (z_n - \xi^3)^2} \chi$$

$$J_n(\xi) d\xi \quad (3.1b)$$

The hybrid algorithm derives its speed advantage by using the analog computer to perform most of this computation in a parallel and continuous manner.

3.1 THEORY

This section describes the mathematical basis for the hybrid algorithm and its architectures. Particular attention is given to the problems associated with: (1) partitioning the computational tasks, (2) analog and digital representation of variables, (3) analog computer techniques for exponential teneration, and (4) convolution on the analog computer. For convenience, two assumptions are made:

- (1) All analog and hybrid components are perfect, i.e., noise, drift static and dynamic errors do not occur. The effects of these errors are treated in Section 4 below.
- (2) Analog computer components are represented by their block structural continuous system simulation language counterparts, which do not invert and which do not require amplitude or time scaling.

3.1.1 Partitioning of the Computational Tasks

Before the computational tasks can be assigned to either the digital or the analog computer, a number of important properties of equation 3.1 must be examined:

- (1) The integrand of the discrete convolution, 3.1a, may be partitioned into three pieces:
 - (a) $Y_D(y_I, \xi_j) = e^{-\frac{1}{2q} (y_I - a\xi_j)^2}$
 - (b) $Z_D(\xi_j) = e^{-\frac{1}{2r} (z_n - \xi_j^3)^2}$
 - (c) $J_n(\xi_j)$

the first of which depends on y_I while the others do not.

- (2) A similar partitioning may be made for the continuous convolution, equation 3.1b:

$$(a) Y_C(y, \xi) = e^{-\frac{1}{2q}(y-a\xi)^2}$$

$$(b) Z_C(\xi) = e^{-\frac{1}{2r}(z_n - \xi^3)^2}$$

$$(c) J_n(\xi)$$

Again, only the first component depends on y .

- (3) If r and q are of the same order of magnitude, then the exponential, $Z_C(\xi)$, will vary sharply when $|\xi^3 - z_n|$ is small. This will not be true of $Y_C(\xi)$ if $a \leq 1$.
- (4) $J_n(\xi)$ is the result of an arithmetic computation at the conclusion of the preceding iteration, or an initial condition. Further $J_n(\xi)$ is stored in tabular form on a grid which will not, in general, be suitable for the next iterative cycle. The values of $J_n(\xi)$ may be obtained by interpolation (or occasionally extrapolation) on the tabular data.

These observations suggest the following partitioning of tasks:

- (1) Use the digital computer to compute and store the terms $Z_D(\xi_j) J_n(\xi_j)$ for $j = 1, 2, \dots, M$.
- (2) Use M similar analog computer circuits to generate the M functions $Y_C(y_I, \xi)$ where $I = 1, 2, \dots, M$ in parallel, where ξ becomes a continuous variable corresponding to $\{\xi_j\}_{j=1,2,\dots,M}$.
- (3) Transmit $Z_D(\xi_j) J_n(\xi_j)$ from the digital computer to the analog computer maintaining the correspondence between ξ_j and ξ .

- (4) On the analog computer multiply the M $Y_C(y_I, \xi)$ terms by a piecewise continuous function corresponding to $Z_D(\xi_j) J_n(\xi_j)$.
- (5) On the analog computer perform a parallel continuous integration of the product obtained in (4) to obtain M unnormalized points of the new conditional density function, $J_{n+1}^u(y)$.
- (6) When ξ (on the analog computer) has covered an interval corresponding to the domain of $Z_D(\xi_j) J_n(\xi_j)$, halt the analog computation and read the outputs of the M integrators, described in (5) above, into the digital computer.
- (7) In the digital computer, normalize and store the new discrete conditional density function $J_{n+1}(y)$ and compute its mean \hat{x}_{n+1} and variance $\hat{\sigma}_{n+1}$.

For the case of Gaussian plant noise, the analog computer's tasks may be separated into three pieces: exponential generation, integrand formation, and integration. The relationships between these three pieces are shown in Figure 6.

3.1.2 Analog Computer Representation of Variables

The representation of $J_n(\xi_j)$ as a set of discrete pairs was described in Section 2, above, and is illustrated in Figure 1. The function $Z_D(\xi_j)$ is stored similarly. $Y_C(y, \xi)$ is, of course, an infinite set of functions, that is, there is a function $Y_C(y_I, \cdot)$ for every fixed value $y = y_I$. M of these functions, $Y_C(y_I, \xi)_{I=1,2,\dots,M}$, are generated as continuous functions of time on the analog computer. Four typical $Y_C(y_I, \xi)$ terms are shown in Figure 7. In order to form an integrand corresponding to the discrete terms of (3.1a) $\{Y_D(y_I, \xi_j) \cdot Z_D(\xi_j) \cdot J_n(\xi_j)\}_{j=1,2,\dots,M}$ for $I = 1, 2, \dots, M$ or to the continuous terms of (3.1b) $Y_C(y, \xi) Z_C(\xi) J_n(\xi)$ the digitally stored terms

must appear in the analog computer as functions of time. Define

$$D_j = \{[Z_D(\xi_j) \cdot J_n(\xi_j)], \xi_j\} \quad j = 1, 2, \dots, M \quad (3.2)$$

to be a set of discrete pairs stored in tabular form in the digital computer. D_j may be converted into a piecewise continuous function of t by introducing

$$Z'_{PC}(t) = \sum_{j=1}^M Z_D(\xi_j) J_n(\xi_j) [u(t-t_1) - u(t-t_2)] \quad (3.3)$$

$$\text{where } t_1 \leq t < t_2$$

$$\text{and } t_1 = \Psi_n(\xi_j)$$

$$t_2 = \Psi_n(\xi_{j+1})$$

where the digital representation's running variable ξ_j is mapped into the analog computer's independent variable time, t , by the mapping

$$\begin{matrix} \Psi_n \\ \xi \longrightarrow t \end{matrix}$$

which consists of a scaling and translation. That is:

$$t = \Psi'_n(\xi) = b_n(\xi + c_n) \quad (3.4a)$$

where

$$b_n = t/[2N\hat{\sigma}_{n-1} + 2N\hat{\sigma}_{n-1}/(M-1)] \quad (3.4b)$$

$$c_n = N\hat{\sigma}_{n-1} - \hat{x}_{n-1} \quad (3.4c)$$

For convenience below, the subscripts on b and c will be omitted, and it will be understood that b and c are functions of the step number, n . A typical $Z'_{PC}(t)$ resulting from this mapping is seen in Figure 8. The output process of the digital computer is characterizable as a zero order hold (ZOH) process. The transfer function of the ZOH may be derived from the Laplace transform

of its impulse response:

$$G_{ZOH}(s) = \frac{1 - e^{-Ts}}{s}$$

If the sampling interval T (seconds/sample) is small compared to the period of the highest frequency in the signals being processed, then it can be shown*

$$G_{ZOH}(s) \cong e^{-Ts/2}$$

Thus, the use of a ZOH reconstruction at the digital-analog interface results in an almost pure delay of $\frac{1}{2}T$. For a more detailed discussion of this, see Miura and Iwata [9]. Since $Z_D(\xi_j)$ is known beforehand, it is easy to translate it forward by this amount, cancelling the delay and eliminating a possible source of errors. This translation is accomplished by representing the set of discrete products D_j on the analog computer by the time function

$$Z_{PC}[t] = \sum_{j=1}^M Z_D(\xi_j) J_n(\xi_j) U(t) \quad (3.5)$$

with

$$U[t] = u(t-t_1) - u(t-t_2)$$

$$\text{where } t_1 \leq t < t_2$$

$$\text{with } t_1 = \Psi_n(\xi_j - \frac{\Delta\xi}{2})$$

$$t_2 = \Psi_n(\xi_j + \frac{\Delta\xi}{2})$$

$$\Delta\xi = \xi_{j+1} - \xi_j$$

as shown in Figure 9 with the map

$$\begin{matrix} \Psi_n \\ \xi \longrightarrow t \end{matrix}$$

That is,

$$t = \Psi_n(\xi) = b(\xi + c) \quad (3.6)$$

*A factor of $1/T$ has been eliminated for convenience. The original sampling process contained a constant, T , which cancels it out.

with

$$b = T / [2N\hat{\sigma}_{n-1} + \frac{2N\hat{\sigma}_{n-1}}{(M-1)}]$$

$$c = [N\hat{\sigma}_{n-1} + \frac{N\hat{\sigma}_{n-1}}{(M-1)}] - \hat{x}_{n-1}$$

where b and c are in general, functions of n .

This transformation considers the probability mass represented in the digital computer by a point mass located at ξ_j to be represented on the analog computer by the area under a step function located on an interval whose midpoint is at $t_j = \psi_n(\xi_j)$.

For future use we add

$$\psi_n^{-1}: \xi(t) = \frac{t}{b} - c \quad (3.7)$$

with b and c as in (3.6).

These definitions differ from those given previously (Eq. 3.4). All subsequent references to ψ_n , b and c will refer to Eqs. 3.5 and 3.6. As a final point, we note that the analog computer's total run time T , which determines the constants in (3.6), is chosen as the smallest value for which accurate hybrid computation can be performed by a given computer system for a particular filter, see Section 4 below.

3.1.3 Exponential Generation

Having determined that the exponentiation required to develop the Y_C terms of Eq. (3.1) will occur on the analog computer and that a suitable mapping of the variable ξ into time t , is given by (3.6), we turn to a discussion of the actual generation:

$$Y_C[y_I, \xi(t)] \triangleq e^{-\frac{1}{2q}[y_I - a\xi(t)]^2} \quad (3.8)$$

with $l = 1, 2, \dots, M$

where $\xi(t)$ is given by (3.7). This is depicted in Section I of Figure 6. For convenience we define

$$e^{W_I(t)} \triangleq e^{-\frac{1}{2q}[y_I - a\xi(t)]^2} \quad (3.9)$$

and

$$W_I(t) \triangleq -\frac{1}{2q}[y_I - a\xi(t)]^2 \quad (3.10)$$

Generalized integration. Generalized integration is a name given to a method often used to generate explicit composite functions on the analog computer (10), (11). This technique requires:

- (1) The derivative of the argument with respect to t be available during the generation of the function.
- (2) The initial condition for the function be supplied for $t = t_0$, where t_0 is the time at which function generation starts.

Let $s(t)$ be the output of an analog integration element:

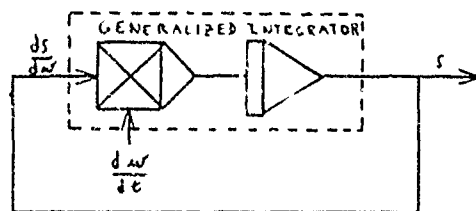
$$s(t) = \int_{t_0}^t (e_1 + e_2) dt$$

we desire $s(t) = e^w$ differentiating once

$\frac{ds}{dw} = e^w$ or $\frac{ds}{dw} - s = 0$. The solution of this auxiliary equation is the desired continuous function. Since M of these functions must be generated simultaneously, and all will depend on their arguments, y_I , it is not possible to establish a correspondence between w and t . In each, generalized integration is involved which depends on the following application of the chain rule for differentiation of composite functions

$$\int \frac{ds}{dw} \cdot dw = \int \frac{ds}{dw} \frac{dw}{dt} dt \quad (3.11)$$

If $\frac{dw}{dt}$ is available in the analog computer, then



Application to the state update exponentials. Applying the method to 3.9

$$\dot{W}_I(t) \triangleq \frac{dW_I(t)}{dt} = \frac{dW_I[\xi(t)]}{d\xi} \cdot \frac{d\xi(t)}{dt} \quad (3.12)$$

from (3.7), $\frac{d\xi}{dt} = \frac{1}{b}$

$$\dot{W}_I(t) = \frac{a}{bq} [y_I - a\xi] \quad (3.13)$$

This can be expanded to

$$\dot{W}_I(t) = \frac{a}{bq} [y_I - a(\frac{t}{b} - c)] \quad (3.14)$$

As mentioned above, b and c are functions of n , as are the y_I . To obtain the initial condition

$$e^{W_I[\xi(t_0)]} \quad \text{expand (3.9) using (3.7) with } t = t_0$$

$$e^{W_I[\xi(t_0)]} = e^{-\frac{1}{2q} [y_I - a(\frac{t_0}{b} - c)]^2} \quad (3.15)$$

For the case $t_0 = 0$ this reduces to

$$e^{W_I[\xi(0)]} = e^{-\frac{1}{2q} [y_I - ac]^2} \quad (3.16)$$

Hardware constraints often preclude starting the generation of the e^{W_I} terms at $t_0 = 0$. This is discussed in Section 3.2.2. The derivative terms $\dot{W}_I(t)$, $I = 1, 2, \dots, M$ shown in Section I of Figure 6 are also generated by the analog computer. The constants necessary for production of these derivative terms, as well as the initial conditions, $e^{W_I[\xi(t_0)]}$, are sent from the digital computer prior to starting the analog solution. See the discussion in Section 3.2 for additional implementation details.

3.1.4 Integrand Formation

Returning to Figure 6, we note that the M terms $\{e^{W_I(t)}\}_{I=1,2,\dots,M}$ generated in Section I are each multiplied in Section II by the same term. The output of the digital to analog converter (DAC) in Figure 6 is set to the proper amplitude every Δt seconds under digital computer timing control.

The change in representation from the set of M discrete pairs D_j stored in the digital computer to the piecewise continuous function $Z_{PC}[t]$ during a run on the analog computer is based on the map y_n and constants b and c which are derived from the conditional mean and variance estimates of the previous iterative cycle.

Subsequent sections discuss the manner in which this digital to analog data transmission is performed and some of the problems and limitations associated with it. Assuming $Z_{PC}(t)$ of proper form, the analog computer performs M parallel continuous multiplications as shown in Section II of Figure 6.

3.1.5 Integration

The last step required by the convolution of (3.1) is the M parallel integrations of

$\{e^{W_I(t)} Z_{PC}(t)\}_{I=1,2,\dots,M}$. The integrations proceed as indicated in Section III of Figure 6, continuously, in parallel and simultaneous with the generation of the $e^{W_I(t)}$ terms and the injection of $Z_{PC}(t)$ which (see Table I) is identical to $J_n|n(t)$.

If the observation, z_n , is too far from \hat{x}_n , or r is too small, severe scaling problems will occur in outputting $Z_{PC}(t)$ to the analog computer. These rather unlikely events are treated in Section 3.2.2 below. We note also in Section III of Figure 6, that in addition to

$$J_{n+1}^u(y_I) = \int_0^T e^{W_I} Z_{PC}(t) dt \quad (3.18)$$

the analog computer is able to generate

$$y_I J_{n+1}^u(y_I) = \int_0^T y_I e^{W_I} Z_{PC}(t) dt \quad (3.19)$$

and

$$y_I^2 J_{n+1}^u(y_I) = \int_0^T y_I^2 e^{W_I} Z_{PC}(t) dt \quad (3.20)$$

for $I = 1, 2, \dots, M$

at the expense of some additional equipment. The judgment on whether to do this depends considerably on the computing power of the digital computer and on the analog equipment available, see Section for additional discussion.

3.2 REALIZATION OF THE HYBRID ALGORITHM

There are two steps required to produce an operating hybrid optimal filter: (1) definition of the equations to be solved by each computer, and (2) development of software to link the two computers together realistically. In the process of partitioning, the first requirement was satisfied. The reader familiar with hybrid computation will recognize the myriad of details which must be carefully treated in linking an analog and digital computer together.

A flow chart of the hybrid algorithm (for the cubed sensor problem) is given in Figure 10. The information paths within the hybrid computer systems are shown in Figure 11. The analog computer diagrams for the convolution integral are presented in Figures 12 and 13.

3.2.1 Information Flow in the Hybrid Filter

The hybrid filter is best described by progressing through the flow chart of Figure 10, with occasional references to the flow charts of the purely digital 1-step predictor, Figures 2, 3 and 4. The hybrid flow chart, Figure 10, assumes a single random sequence for its input. In general these will be noisy observations of an on-line stochastic system. The functions performed within each box shown in Figure 10 are:

BOX 1. The digital computer initializes parameters: q - plant noise variance; r - observation variance; k - initial state estimate variance; M - grid discretization; N - number of estimated standard deviations, $\hat{\sigma}_{n-1}$, about the $(n-1)$ st

estimated mean, \hat{x}_{n-1} , spanned by the n^{th} grid (see Figure 1)*; and J_0 - the assumed initial conditional density function. The zeroth grid base is laid out using J_0 as a reference. The observation number, n , is set to -1. The processing performed here is identical to that of BOX 1 in the all digital filter algorithm of Figure 2. The digital computer calculates and transmits to the analog M initial values of y_i - the point at which $J_{n+1}(y)$ will be computed. The initial conditions

$W_I(t_0)$ are stored in the analog computer's sample and hold elements, as are the initial values of the scaling constant, b_0 , and the translation constant c_0 . Because of analog computer limited dynamic range (see Section 3.2.2), the initial conditions will be too small to be adequately set on the analog computer at time $t = 0$. The digital computer determines the time in the solution, t_{01} , at which each of the M exponential generators can be accurately started. When this time is reached, the analog gate shown in Figure 12 will be closed and the exponential generation $e^{W_I(t)}$ begun.

BOX 2. The sample sequence number n is incremented. All iterative cycles begin here.

BOX 3. For a simulated system the digital computer performs a model and density update identical to that represented by BOX 3 of Figure 2. In the case of an on-line stochastic input, past data corresponding to the incremented count is assumed to be available. However, it need not actually arrive until required by the computation at BOX 7.

BOX 4. The analog computer is placed in the initial condition (IC) mode. Initial conditions are set into the analog computer integrators. The analog gates in Figure 12 are closed only for those integrators at which $t_{01} = 0$. Other circuit configurations may be more desirable. The important

*As indicated in the discussion at BOX 5, the grid base is not changed between iterations unless the computation requires it. Hence, in general, N is the number of estimated standard deviations, $\hat{\sigma}_{n-r}$, about the $(n-r)^{\text{th}}$ estimated mean \hat{x}_{n-r} spanned by the n^{th} grid where r is the number of times since the grid was last changed.

point is that limited analog computer dynamic range requires that some integrators start "later" than others because their initial value does not reach an accurately settable value until $t \geq t_{0I}$ where t_{0I} is a grid base and problem dependent number (see Section 3.2.2).

BOX 5. A check is made to determine whether the grid should be floated, i.e., whether the mesh or location of the grid base on which the $(n+1)$ st probability density function $J_{n+1}(y)$ is to be constructed should be changed from that on which the n^{th} probability density function, $J_n(\xi)$, was placed. Although much time can be saved in both purely digital and hybrid filtering if the grid base does not change, the percentage savings in hybrid computation are significantly greater. One must, of course, be very careful that the grid mesh is not mislocated or too narrow, causing significant amounts of "probability mass" to be located off the end of the grid. Likewise, a grid which is too coarse will not be able to properly resolve the probability mass. BOX 5 compares \hat{x}_n with \hat{x}_{n-r} and $\hat{\sigma}_n$ with $\hat{\sigma}_{n-r}$, where r is the number of observations processed since the last grid float, and branches to BOX 6 if either of the differences is too great.

BOX 6. The grid is floated by recomputing the scale factor, b , and translation factor, c , followed by computing M new values for y_I and $W_I(t_0)$. This information is DAC'd to the analog computer. M new values of t_{0I} , the time of application of the 1^{th} initial condition $e^{W_I(t_0)}$ to the 1^{th} exponential generator, are computed and stored. Interpolation and possibly extrapolation are required for $J_n[\xi(j)]$ because values of $\xi(j)$ at which its M constituent delta functions are located should correspond with the new values for y_I . For a "smooth" $J_n[\xi]$ a second set of interpolations is required to produce $(k-1) \times M$ intermediate values. The $k \times M$ discrete pairs representing $J_n(\xi)$,

$\{J_n[\xi(l)], \xi(l)\}_{l=1,2,\dots,M}$, are stored for future use during the current analog solution as well as for all future solutions on the same grid base. For interpolated values, $J_n(\xi_l)$ and $Z_D(\xi_l)$, define a set of discrete pairs

$$\begin{aligned} D_l &= \{[J_n(\xi_l)], \xi_l\}_{l=1,2,\dots,k \times M} \\ &= \{[Z_D(\xi_l)J_n(\xi_l)], \xi_l\}_{l=1,2,\dots,k \times M} \end{aligned} \quad (3.2a)$$

stored in the digital computer corresponding to D_j (see Eq. 3.2). If k is chosen odd, the mapping (3.6) may be maintained and the time function appearing on the analog will be

$$Z_{PC}^{(l)}(t) = \sum_{l=1}^{k \times M} Z_D(\xi_l) J_n(\xi_l) U_1(t) \quad (3.5a)$$

where U_1 is defined as in 3.5, but with running subscript l . As $Z_{PC}^{(l)}(t)$ is identical in form to $Z_{PC}(t)$ shown in Figure 9 (but with increased discretization), the superscript will be suppressed in this sequel. For this kind of intermittent grid float, the span and center of ξ and y are made identical in order that each succeeding $J_{n+1}(y)$ is developed on a grid base suitable for its injection as $J_n(\xi)$ on the next iteration.

BOX 7. For an on-line system there may be a wait for the n^{th} data sample, z_n . Using z_n , $Z_D(\xi_1)$ is computed. The product $Z_D(\xi_1) \cdot J_n[\xi(1)]$ is formed. Under certain conditions (see Section 3.2.2) those values, for $l > 1$ (corresponding to $t > 0$), may be computed in parallel with the operations performed in BOXES 8-14 as long as they are ready before the analog computer requires them.

BOX 8. The running index, l , of the piecewise continuous function $Z_{PC}(t)$ (see Eq. (3.5a)), is set to 1. $Z_D(\xi_1)J_n(\xi_1)$, or what is equivalent, $J_n|_n(\xi_1)$ is DAC'd from the digital to the analog computer.

BOX 9. The analog computer is placed in the COMPUTE mode. This is time t_0 , or, as is the custom, $t_0 = 0$.

BOX 10. The analog computer simultaneously and continuously generates

$$e^{W_I(t)} = e^{-\frac{1}{\tau q} [y_I - a\xi']^2} \quad \text{for } I = 1 \text{ to } M, \text{ where}$$

$\xi(t) = \frac{t}{b} + c$, see Eq. (3.7) and Figure 7. In addition, it performs a continuous integration of the product $e^{W_I(t)} Z_{PC}(t)$. When $t = \tau[\xi_\ell + \frac{\Delta \xi_\ell}{2}]$,

i.e., the value of t corresponding to the time of injection of the next digitally stored value of $Z_{PC}(t)$, BOX 11 is entered.

BOX 11. A test is made to determine whether the $(k \times M)^{\text{th}}$ value of $J_n(\xi_\ell)$ has been processed.

BOX 12. If $\ell < k \times M$ the index ℓ is incremented by one.

BOX 13. The next digitally stored value of $J_n(\xi)$, $J_n(\xi_\ell)$ is DAC'd to the analog computer.

The analog computer remains in COMPUTE while ℓ is incremented from 1 to $k \times M$. During this time, it integrates

$(e^{W_I(t)} \cdot e^{W_Z(t)} \cdot J_n(t))_{I=1, 2, \dots, M}$. This operation corresponds to the one taking place in BOX 4 of Figure 2, labeled Integral Evaluation. BOX 4 of Figure 2 is expanded in Figure 3. It is the difference in nature of the computation being performed by the hybrid filter in BOXes 9-13 of Figure 12, from that being performed by the digital filter in BOX 4 of Figure 2, that leads to the major differences in speed and accuracy between the two optimal filters.

BOX 14. If the answer to the question posed in BOX 11 is positive indicating that $k \times M$ time sub-intervals have elapsed (or equivalently that $t = T$), the analog computer is put into HOLD (i.e., integrator inputs are disconnected).

BOX 15. The M unnormalized $(n+1)$ st probability density function points, $J_{n+1}(y_I)$, are read from analog to digital converters (ADC'd).

BOX 16. The digital computer performs normalization to produce $J_{n+1}(y)$, followed by computation of \hat{x}_{n+1} and \hat{q}_{n+1} . This calculation is identical to that performed in BOX 5 of Figure 2 and expanded in Figure 4.

BOX 17. A check is made to see if the last piece of data, $z_{n_{\text{MAX}}}$, has been processed. If $n < n_{\text{MAX}}$, n is incremented (BOX 2) and another iteration of the algorithm occurs. If $n = n_{\text{MAX}}$, the exit is taken.

3.2.2 Further Details of the Hybrid Algorithm

The discussion below provides additional insight into the functioning of the hybrid algorithm. Analog and hybrid errors are treated briefly.

Grid floating. If the plant is unstable ($a > 1$ in Eq. 2.1), or if successive estimates do not converge in phase space, then it is difficult to discretize the conditional density function $J_{n+1}(y)$. If the range on y is made large, then an unreasonable number of discretization intervals results. If the discretization is made coarser, then the accuracy is impaired. The solution to the impasse is to float the grid (see References 3 and 4). The floated grid permits the accuracy of a finely discretized grid to be obtained with a minimum of computer time, by placing the discretization where it will do the most good. In the purely digital

computation if the grid were fixed (i.e., not floated), then the computations described above could be simplified considerably. The functions

$$e^{-\frac{1}{2q}(y_I - \xi_j)^2}$$

$$e^{-\frac{1}{2r}(z_n - \xi_j^3)^2}$$

could be precomputed and stored in large arrays. Some interpolation will be required for the function which contains z_n and the arrays may be quite large. Equivalently, in the hybrid algorithm, the exponential generators, $Y_C(y_I, \xi)$, could be established for the entire observation sequence. The sample and hold elements shown at the output of the digital to analog converters (DAC's) in Figure 13 would not be necessary. When the range of y in $J_{n+1}(y)$ is known in advance and is not too large, acceptable accuracy has been obtained using both floated and nonfloated grids. As the range of y increases, the difference in computation time for each estimate on the floated and nonfloated grid becomes larger, at equivalent accuracy. However, those problems which can be represented by a cyclical, or modulo- 2π density function not requiring high local resolution such as the phase lock loop error, see (12), can be represented on a fixed grid.

Filter errors at low data rates. Analog computer components are subject to a variety of low frequency and static errors

- (a) noise
- (b) offset
- (c) drift
- (d) nonlinear behavior

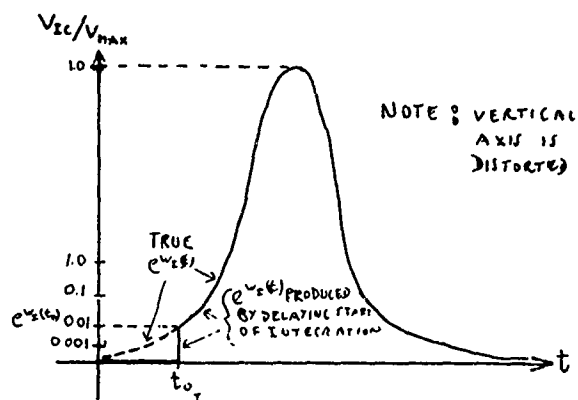
Contemporary, 3rd generation analog computers (19) are available with low frequency and static accuracies on the order of 0.01%. As the number of components used in a problem increases, these errors are compounded. Further, the range of useful computation on such analog computers is ± 10 volts. If the noise level is approximately 1 to 10 millivolts, then the useable dynamic range of an analog computer is 1 part in 10^4 or 10^3 .

This limited dynamic range should be contrasted with 1 part in 10^{150} available in the IBM 360 series machines, and the even greater dynamic range available from CDC computers. The density function of a normally distributed random variable drops three decades in 3.2 standard deviations either side of the mean. It is not uncommon for an optimal filter to span 12 or 16 standard deviations. Digital computers easily generate density functions spanning 72 standard deviations (± 36 deviations on either side of the mean). Consider the exponential generator in Figure 6.

To generate a "bell" on the analog computer, an initial condition is needed which corresponds to a point on one of its "skirts". If we were to attempt to set all initial conditions at the beginning of an analog computer run for the 1-D cubed sensor problem with $a = 0.5$, fully one half of all these initial values correspond to points three or more standard deviations from the mean. That is, half of all initial conditions would have to be set to an accuracy below the analog computer's dynamic range. To avoid large errors in the bells, the start of integration is delayed until $Y_C(y_I, \xi)$ reaches a value which can be accurately set at the integrator's output, approximately 10 MV in a 10 V full scale analog computer. This value is stored in the sample and hold device until the time corresponding to the initial condition is reached, see Figure 12. The time of application t_0 of an IC is calculated using Eq. 3.15 with

$$e^{W_1[\xi(t_0)]} = .001.$$

A signal $\lambda_{\xi_i}(t)$ from the i^{th} digital output point (DO) in the discrete data control unit (see Figure 11) releases the integrator. If all of the decades of precision were visible in a graph, then the exponential function and its corresponding analog computer equivalent might appear as follows:



In Section 5 we report results of a hybrid optimal filter simulation in which the ratio of V_{IC} to V_{MAX} is varied. At the output circuit the principle source of low frequency errors is located at the integrator input where small observation noise variance or a data point z_n corresponding to an x_n far from \hat{x}_n lead to a needlelike $J_n|n(t)$. Multipliers are built such that they produce full scale output when full scale input is applied to both inputs. In terms of threshold voltages (V_{TH}), if the full scale voltage is $10^4 V_{TH}$ then the product of two inputs, each 100 times V_{TH} , produces only a threshold voltage output. Since contemporary integrators have excellent equivalent input drift and noise characteristics, the primary problem encountered in this circuit is to obtain multiplier outputs which are above V_{TH} for significant inputs. Assuming that the exponential generators are always set to achieve full scale voltage at $a\xi = y_1$, only the scaling or method of multiplying by $J_n|n(t)$ may be varied, to avoid loss of accuracy. The set D_ℓ can be totally precomputed and rescaled (at BOX 7 of Figure 10) or alternatively multiplying DACs instead of multipliers can be used for output multiplication. Both slow the computation down and the latter increases cost as well. It is also possible to bypass the analog computer and perform a simple analytical computation corresponding to a stochastic state update for the remote data point case. See (13) for a discussion of the case where $x_n \approx z_n^{1/3}$.

Filter Errors Associated with High Speed Hybrid Computation. Having discussed some of the

problems which arise if we can assume the analog computer has an amplitude accuracy of 1 part in 10^3 or 10^4 , it is important to determine when this assumption can be made. Above a certain data rate, finite analog computer component bandwidths and nonzero data transmission and mode switching times will so deteriorate analog computer performance as to invalidate accuracy analysis based on purely static errors. Component bandwidth limits the fidelity with which the exponential generators can produce $Y_C(y_1, \xi)$ and the output multipliers can produce

$$e^{W_I(t)} J_n|n(t).$$

In Section 4 we present analytical results in which the exponential integrator is replaced by a second order system representing its high frequency equivalent circuit. The output multiplier again may be approached more directly. In Section 4 we present analytical results considering two limiting cases for the form of $J_n|n(t)$.

Intercomputer maximum data rates and hybrid system switching times are a second source of errors in high frequency hybrid filtering. The hybrid optimal filter has the advantage that for low dimensional problems, i.e., $d \leq 3$, $J_n|n(\xi_\ell)$ may be computed in its entirety by the digital computer prior to convolution on the analog computer. That is, the two computers can operate essentially sequentially. This can be seen in Figure 10 where for $n \geq 0$ the digital computer performs computations during BOXes 16, 17, 2, 3, 5, (6), and 7 followed by the analog performing the computation in BOX 10. The linkage system and the digital computer do not have to work very hard while the analog computer solves simultaneous differential equations at high speed. This situation does not prevail in closed loop hybrid computation. Maximum channel data transfer rates place an obvious upper bound on filter data rate. For example, to produce $J_n^u|n(y)$ on the analog computer in 1 MS with $\ell = 100$ $J_n|n(\xi_\ell)$ must be transmitted at a 100 KH rate. That is, every 10 μs , a word located in computer memory must be transmitted to a register at an interface (see Figure 11) and converted to an

equivalent stable analog voltage available at the DAC output shown in Figures 6 and 12. Although higher speed equipment is available, maximum channel data rates appear to set 1 MS as a lower bound for a single convolution on the analog if $Z_{PC}(t)$ is of the form shown in Figure 9.

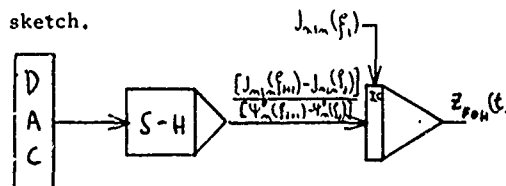
Even though the analog computer performs the convolution in an essentially open loop manner, at high repetition rates, phase shifts caused by timing imprecision (i.e., earliness or lateness) in operation of integrators, sample and hold mode control switches, and analog gates are an additional cause of inaccuracy.* Specifically (see Figures 11, 12, and 13), these errors are caused by the limits on timing precision that the $\lambda_{E_I}(t)$ signals can close the analog gates, and the high speed

DAC can supply $J_n(t) e^{W_Z(t)}$ with respect to each other and with respect to the time origin set by the COMPUTE mode control switch which starts the time base integrator. We must add to this timing skew and the effects of deteriorated signal waveforms of the pulses transmitted through the linkage system to initiate these operations. It can be shown, that a timing imprecision of only 16 nanoseconds produces a .01% F.S. amplitude error for a sine wave with 1 MS period. We defer detailed examination of this point until Section 4 and mention only that timing imprecision is a not well advertised hybrid computer characteristic and that a 300 nanosecond switching grey band is a more reasonable value to expect for contemporary analog computers with 900 ns - 1 μ s total switching times.

Interpolation of $J_n|_n(\xi)$. As a final detail we discuss the advantages and difficulties of injected function interpolation. First note that no matter how it is represented in the digital or analog computers $Z_D(\xi)J_n(\xi)$ or $J_n|_n(\xi)$ is a continuous function. The $Y_C(y_I, \xi)$ terms generated on the analog

are also continuous. Therefore interpolation of $J_n(\xi_j)$, given accurate values at ξ_j will increase the accuracy of computation. Results given in Section 5 demonstrate this. However interpolation lengthens digital computation time and overlapping the computation leads to output scaling problems discussed above. In addition, as mentioned, the greater the number, $l = k \times M$, the more data which must be transferred between computers per analog run and the larger the minimum T as determined by maximum channel data rate must be. Another approach is the use of a first order hold FOH. This device (see (16)) essentially performs linear interpolation on the data after transmission

It does not require a $\frac{\Delta \xi_j}{2}$ advance and requires less transmitted data points for equivalent amplitude accuracy. Only a simple circuit is required, see sketch.



FIRST ORDER HOLD

4. THE EFFECTS OF HIGH DATA RATES ON HYBRID FILTER ACCURACY

As mentioned, there is some data rate, above which, analog computer static errors are no longer reliable guides to solution accuracy. In this section we consider for the scalar filter just what the tradeoffs are. Again the cubed sensor problem serves as an example.

4.1 ANALOG COMPUTER COMPONENT BANDWIDTH

4.1.1 Exponential Generator

Under certain reasonable assumptions an operation amplifier can be represented by the following

*For a discussion of this point see (14) and (15), where the nature and effects of H. Y. Bridg's law also called the "Magic Equation" are discussed.

transfer function

$$G(s) = -\frac{z_f}{z_i} \cdot \frac{1}{1 + \frac{1}{A\beta}} \quad (4.1)$$

where A = open loop amplifier GAIN

$$\beta = z_i / (z_i + z_f)$$

z_f = feedback network short circuit transfer impedance

z_i = input network short circuit transfer impedance

For $A \gg 1$, the usual approximation

$$G(s) \approx -\frac{z_f}{z_i}$$

results. At high frequencies it is necessary to replace A with $A(j\omega)$ to evaluate operational amplifier performance. $A(j\omega)$ is, in general, a complex transfer function with 3 or more break points. Simplified analysis valid for some operational amplifiers assumes a single pole transfer function

$$\frac{A_0}{\tau_i s + 1} \quad (4.2)$$

Substitution of 4.2 into 4.1 and setting $\frac{z_f}{z_i} = \frac{1}{\tau_i s}$ (WHERE $z_f = \frac{1}{C_s} \Rightarrow \tau_i = RC$)

for an integrator yields a second order transfer function containing two widely separated poles. The low frequency pole represents the physical fact that the amplifier gain is not infinite at DC.

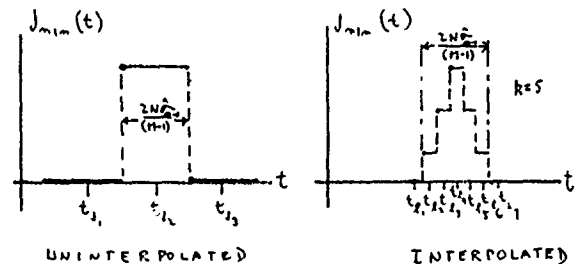
The location of the high frequency pole determines the effect of limited bandwidth on hybrid filter performance. For $A_0 = 10^8$, $\tau_i = 1/2000$, $\tau_i = 10$, $N = 8$, $T = 4$ MS, $\hat{\omega}_z \approx 8$, $\hat{\omega}_n \approx 0$

It can be shown that the highest frequency of interest in $\omega e^{\omega_z t}$ will not produce an error in excess of .01%. Since the exponential generators are nonlinear, this analysis can only be considered a rough guide. See (26) for results of a simulation of amplifier limited bandwidth in an exponential generator.

4.1.2 Output Multiplier

At the output multiplier $e^{\omega_z t}$ is multiplied by $J_{min}(t)$ which has two possible worst case con-

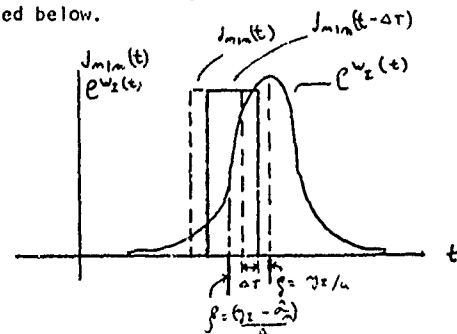
figurations



If we assume the fundamental and third harmonic contain all terms contributing significant mass to $J_{m+1}(\gamma_z)$ and use manufacturer's multiplier specifications, a 1 MS convolution time, T , leads to errors less than .02%. Results of a simulation of this portion of the circuitry will also appear in (26).

4.2 HYBRID SYSTEM TIMING IMPRECISION

This limitation may be examined by considering the mass shift in $J_{m+1}(\gamma)$ caused by the worst case sketched below.



The boxcar function corresponds to the uninterpolated $J_{min}(t)$ worst case considered above; the bell is $e^{\omega_z t}$. $d e^{\omega_z t} / d \hat{\omega}_z$ can be obtained from normalized tables. For the cucoed sensor problem with $T = 4$ MS, $A = 0.5$, $\hat{\omega}_z \approx 8$, $\Delta\tau$ in sketch above, if 300 nanoseconds (see section 3.2.2) we again find that the errors in $J_{m+1}(\gamma_z)$ are less than .02%.

A similar sort of analysis in which an ideal $J_{m+1}(\gamma_z)$ is compared against the $J_{m+1}(\gamma_z)$ produced when $\lambda_z(t)$'s delayed by 300 NS also yields errors in $J_{m+1}(\gamma_z)$ less than .02%.

5. SIMULATION RESULTS

A simulation of the hybrid algorithm for optimal discrete one step prediction using a modified version of MOESSL, a block structured continuous system simulation language developed at the University of Southern California (17), (18) is underway. The modifications to MOESSL provide an ability to simulate analog computer limited dynamic range and make double precision arithmetic available for highly accurate simulation. Two types of results are available at present.

5.1 VERIFICATION OF THE ALGORITHM

Table 2 compares for 6 data points the estimated means and variances produced by a simulated hybrid filter against results obtained for the all digital filter on the cubed sensor problem. For both filters: $k = q = r = 1.0$, $a = 0.5$ and $N = 8$. For the hybrid filter $M = 29$ and $k \times M = 319$. A dynamic range of 10^{11} was used in this simulation. Digital filter discretizations were $M = 29$ and $M = 513$. Digital filter runs of the form described by Bucy (13) to obtain $\hat{x}_n(M)$ and $\hat{\sigma}_n^2(M)$, have shown that for this problem, $M = 513$ is a sufficiently great discretization to produce estimates which do not differ in the 4th place past the decimal point. Note that the hybrid filter estimates are closer to the highly discretized digital filter estimates than those of the digital filter with $M = 29$.

5.2 EFFECTS OF ANALOG COMPUTER LIMITED DYNAMIC RANGE

Results from a simulation in which analog computer dynamic range was varied are presented in Table 3. Mean and variance estimates obtained from the hybrid filter and a highly discretized all digital filter for three data points are shown. The ratio of V_{MAX} to V_{TH} is varied between 10^2 and 10^{10} . At dynamic ranges of 10^3 or greater the simulated hybrid filter estimates are in close agreement with the purely digital estimates. For both filters: $k = q = r = 1$, $a = 0.5$, $N = 8$. For the all digital filters $M = 513$ and $M = 29$. The hybrid filter grid discretization was $M = 29$ with no interpolation, i.e., $k \times M = 29$.

6. USE OF INTERACTIVE GRAPHICS IN OPTIMAL FILTERING

The USC System Simulation Laboratory contains a Adage AGT-10 graphic computer and IBM-360-44 medium speed scientific computer and a Beckman EasE 2132 Analog Computer. A hybrid linkage connects the computers. Current software permits the graphic computer to be used standalone or in conjunction with the 360 in any of several modes.

At present programs have been written which read a sequence of stored 2-D conditional density function from disk and display them in various ways on the graphic terminal screen. Most of these programs are written in the AGNOS graphic programming language (20).

These display programs can be used to evaluate filter performance by indicating temporary loss of observation data, singularities in the model equations, grid inadequacy, poor modeling, or very poor prior knowledge of state. The display for one of these programs is shown in Figure 14. Any of the 2-D density functions on the periphery may be displayed in the center. A sequence of mounds may be displayed at center, using a "joystick" to control the sequencing rate. The density functions may be blended from one to the next. The graphic output of a program capable of displaying any pre-selected density function is given in Figure 15. The graphic computer screen, its function switches and joystick are foreground. The IBM-360 is in background.

7. SUMMARY AND CONCLUSIONS

The hybrid algorithm for optimal nonlinear discrete filtering has been presented. Timing estimates indicate that the hybrid algorithm will require 10 seconds for one state estimate in a 4-D task. This compares to 3 seconds for 49 CDC 6600's operating in parallel. Thus, one hybrid computer with 250 integrators and multipliers operates at speeds equivalent to 49 CDC 6600's. The relative disadvantage is accuracy. High speed hybrid filter estimates will generally be within 1%.

The effects of analog and hybrid errors are discussed and a more thorough treatment is in progress [Reference 26].

Application areas for the hybrid optimal filter are:

- (1) high data rate, moderate accuracy tasks, i.e., .1% to 1% accuracy is acceptable,
- (2) continued illumination of optimal nonlinear filtering with a view towards more general theory,
- (3) establishing estimates of moderate accuracy as a reference for suboptimal filter estimates in high dimensional tasks.

ACKNOWLEDGMENT

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TABLE 1 HYBRID FILTER NOTATION*

The Bayesian Sequential Estimator:

$J_a|b(\cdot) \triangleq$ Conditional Density of random variable x_a given observations z_0, z_1, \dots, z_b

The following notational simplification is used:

If $a = b + 1$ conditional subscript b is suppressed, e.g., $J_{n+1}(\cdot) \triangleq J_{n+1}|n(\cdot)$

$$J_{n+1}(y_I) = \sum_{j=1}^M Y_D(y_I, \xi_j) Z_D(\xi_j) J_n(\xi_j)$$

$$J_{n+1}(y) = \int_{-\infty}^{\infty} Y_C(y, \xi) Z_1(\xi) J_n(\xi) d\xi$$

- | | | |
|-----|--|---|
| 1 | $Y_C(y, \xi) = e^{-\frac{1}{2q}(y-a\xi)^2}$ | Unnormalized Conditional Density of x_{n+1} given x_n |
| 1.1 | $Y_C(y_I, \xi) = e^{-\frac{1}{2q}(y_I-a\xi)^2}$ | Analog Representation of 1 for fixed $y = y_I$ |
| 1.2 | $Y_D(y_I, \xi_j) = e^{-\frac{1}{2q}(y_I-a\xi_j)^2}$ | Discrete Representation of 1 |
| 2. | $Z_D(\xi) = e^{-\frac{1}{2r}(z_n-\xi)^2}$ | Unnormalized Conditional Density of z_n given x_n |
| 2.1 | $Z_D(\xi_j) = e^{-\frac{1}{2r}(z_n-\xi_j)^2}$ | Discrete Representation of 2 |
| 3 | $J_n(\xi)$ | Conditional Density of Random Variable x_n given Observation z_0, z_1, \dots, z_{n-1} a continuous function |
| 3.1 | $J_n(\xi_j)$ | Discrete Representation of 3 |
| 4 | $D_j = \{[Z_D(\xi_j) \cdot J_n(\xi_j)], \xi_j\}$ | $j = 1, 2, \dots, M$ |
| 4.1 | $D_\ell = \{[Z_D(\xi_\ell) \cdot J_n(\xi_\ell)], \xi_\ell\}$ | $\ell = 1, 2, \dots, k \times M$ |
| 5 | $Z_{PC}(t) = \sum_{j=1}^M Z_D(\xi_j) J_n(\xi_j) U(t)$ | Piecewise Continuous Function Corresponding to 4 |
| | with $U(t) = u(t-t_1) - u(t-t_2)$ where $t_1 \leq t < t_2$ $t_1 = \gamma_n(\xi_j - \frac{\Delta\xi}{2})$, $t_2 = \gamma_n(\xi_j + \frac{\Delta\xi}{2})$, $\Delta\xi = \xi_{j+1} - \xi_j$, γ_n defined in (3.6) | |
| 5.1 | $Z_{PC}^{(k)}(t) = \sum_{\ell=1}^{k \times M} Z_D(\xi_\ell) J_n(\xi_\ell) U(t)$ | Piecewise Continuous Function Corresponding to 4.1 |
| | For k odd define $U(t)$ as in 5 with j replaced by ℓ | |
| 6 | $J_n n(\xi) \triangleq Z_D(\xi) J_n(\xi)$ | (See Reference 8, Eq. 2.5) |
| 6.1 | $J_n n(\xi_j) \triangleq Z_D(\xi_j) J_n(\xi_j)$ | Discrete Representation of 6 |

*Unless otherwise specified in this table or text, the notation is that used in (3), (4) and (8).

7	$e^{W_1(t)} \triangleq Y_C(y_1, \xi)$	Alternate form of 1.1	TABLE 1 (CONTINUED)
8	$e^{W_Z(t)} \triangleq \sum_{j=1}^M Z_D(\xi_j) U(t)$	Piecewise Continuous Function Corresponding to 2.1	
9	$J_n(t) \triangleq \sum_{j=1}^M J_n(\xi_j) U(t)$	Piecewise Continuous Function Corresponding to 3.1	
10	$J_n _n(t) \triangleq \sum_{j=1}^M Z_D(\xi_j) J_n(\xi_j) U(t)$	Piecewise Continuous Function Corresponding to 6.1	
Note: $J_n _n(t) \triangleq Z_{PC}(t)$			
11	$J_{n+1}^u(y_1) = \int_0^T Y_C Z_{PC} dt$	Unnormalized I^{TH} Point of (n+1)st Conditional Density Function $J_{n+1}(y)$	

TABLE 2 COMPARISON OF ESTIMATES FROM TWO ALL DIGITAL OPTIMAL FILTERS WITH THOSE FROM SIMULATION OF AN IDEAL HYBRID OPTIMAL FILTER						
k	ESTIMATED MEANS \hat{x}_{k+1}			ESTIMATED VARIANCES $\hat{\sigma}_{k+1}^2$		
	DIGITAL		HYBRID	DIGITAL		HYBRID
	M = 29	M = 513	M = 29	M = 29	M = 513	M = 29
0	.855	.825	.825	1.005	1.004	1.004
1	-.026	-.022	-.022	1.093	1.092	1.092
2	.334	.333	.333	1.106	1.097	1.096
3	-.110	-.100	-.100	1.099	1.095	1.095
4	.145	.159	.159	1.089	1.096	1.095
5	-.012	-.015	-.015	1.082	1.085	1.084

TABLE 3						EFFECT OF ANALOG COMPUTER LIMITED DYNAMIC RANGE ON HYBRID FILTER ACCURACY						
k	ESTIMATED MEANS \hat{x}_{k+1}						ESTIMATED VARIANCES $\hat{\sigma}_{k+1}^2$					
	DIGITAL			HYBRID			DIGITAL			HYBRID		
	M=513	M=29	$10^2:1$	$10^3:1$	$10^4:1$	$10^{10}:1$	M=513	M=29	$10^2:1$	$10^3:1$	$10^4:1$	$10^{10}:1$
0	.825	.855	.845	.854	.844	.854	1.004	1.001	.869	1.003	1.009	0.989
1	-.022	-.026	-.036	-.039	-.045	-.020	1.092	1.091	1.005	1.090	1.091	1.091
2	.333	.334	.389	.321	.320	.342	1.097	1.106	.979	1.098	1.096	1.104

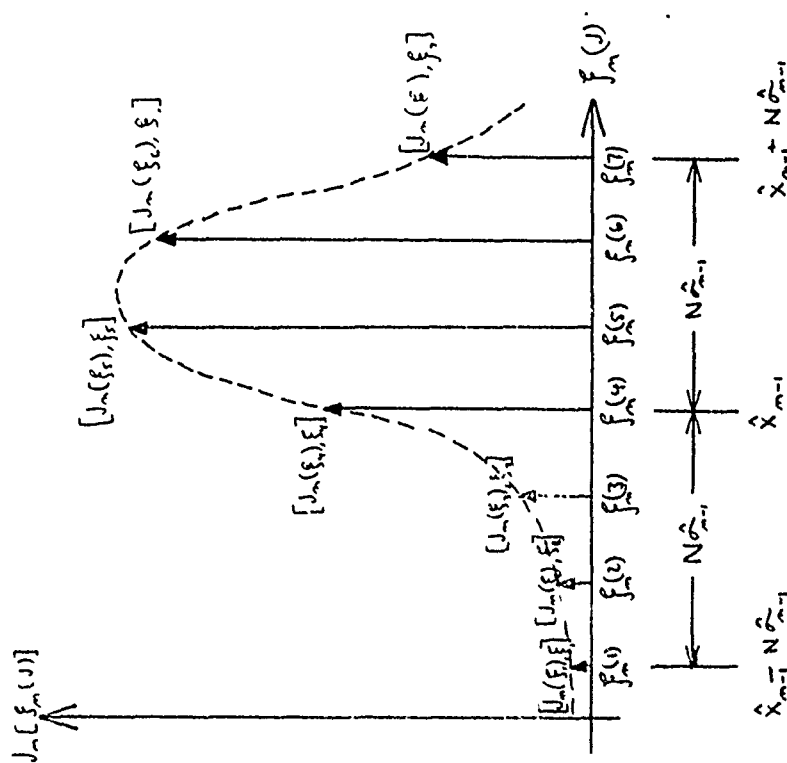


Figure 1. Representation of the n^{th} Conditional Density Function $J_n[f_n(u)]$ by a Set of M Dirac Delta Functions on a Grid Spanning $2N_{n-1}^2$ Centered at f_{n-1} for the Case $M=7$ for a 1-State Variable Problem.

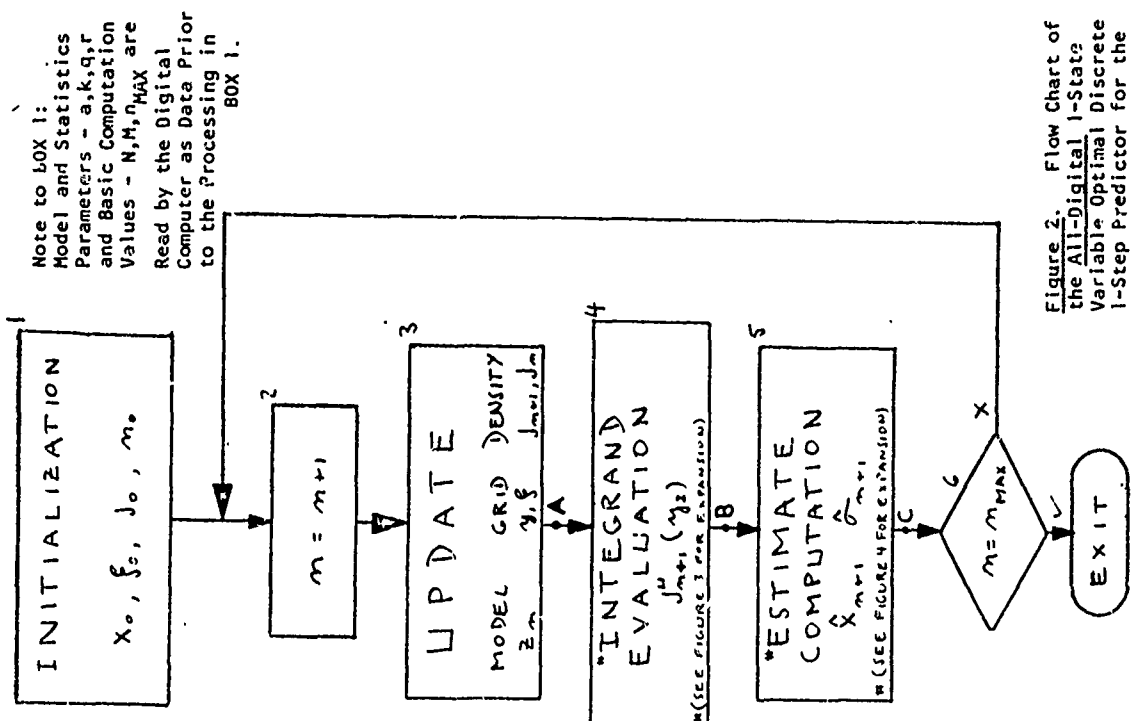
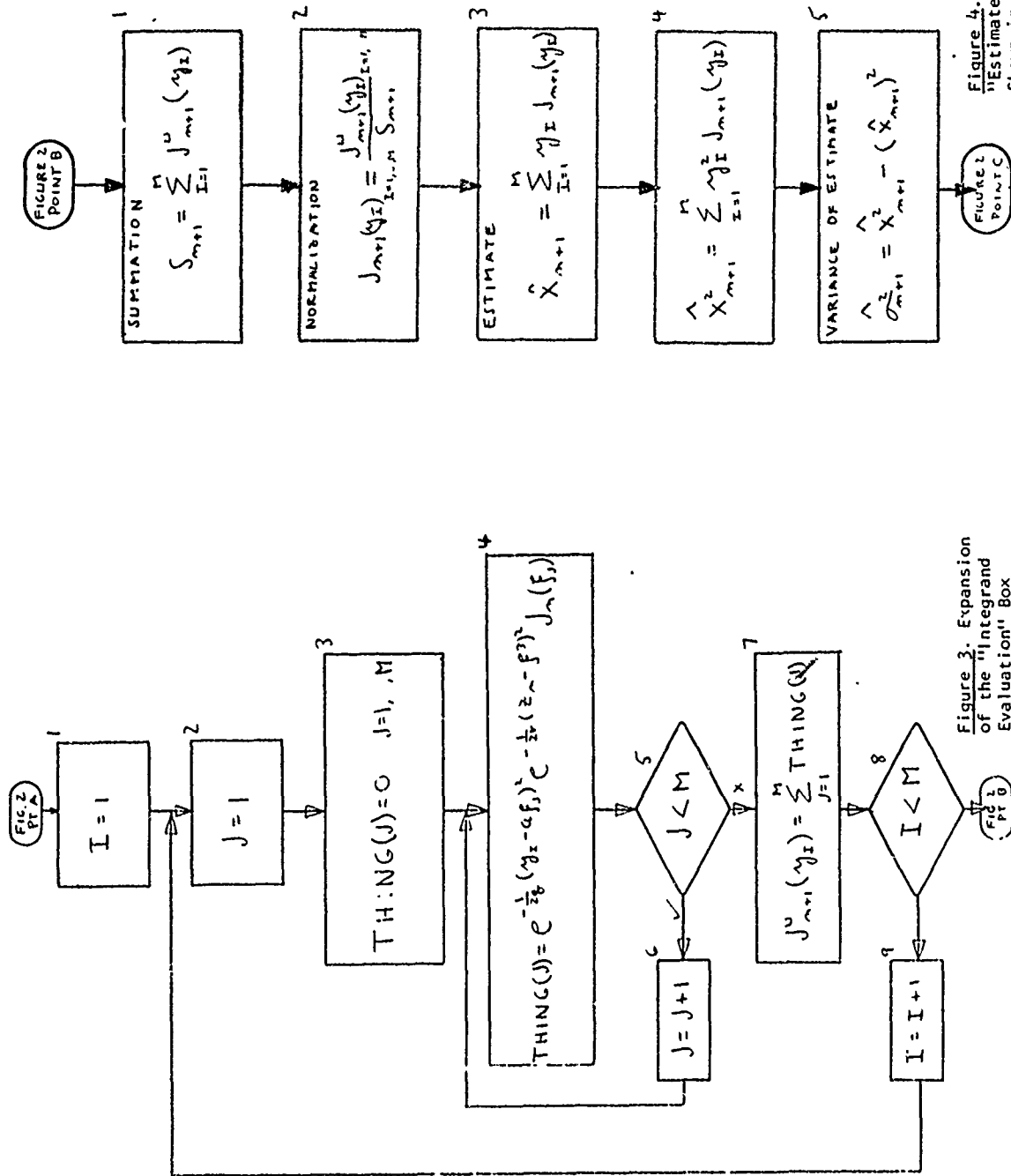


Figure 2. Flow Chart of the All-Digital 1-State Variable Optimal Discrete 1-Step Predictor for the Cubic Sensor Problem.



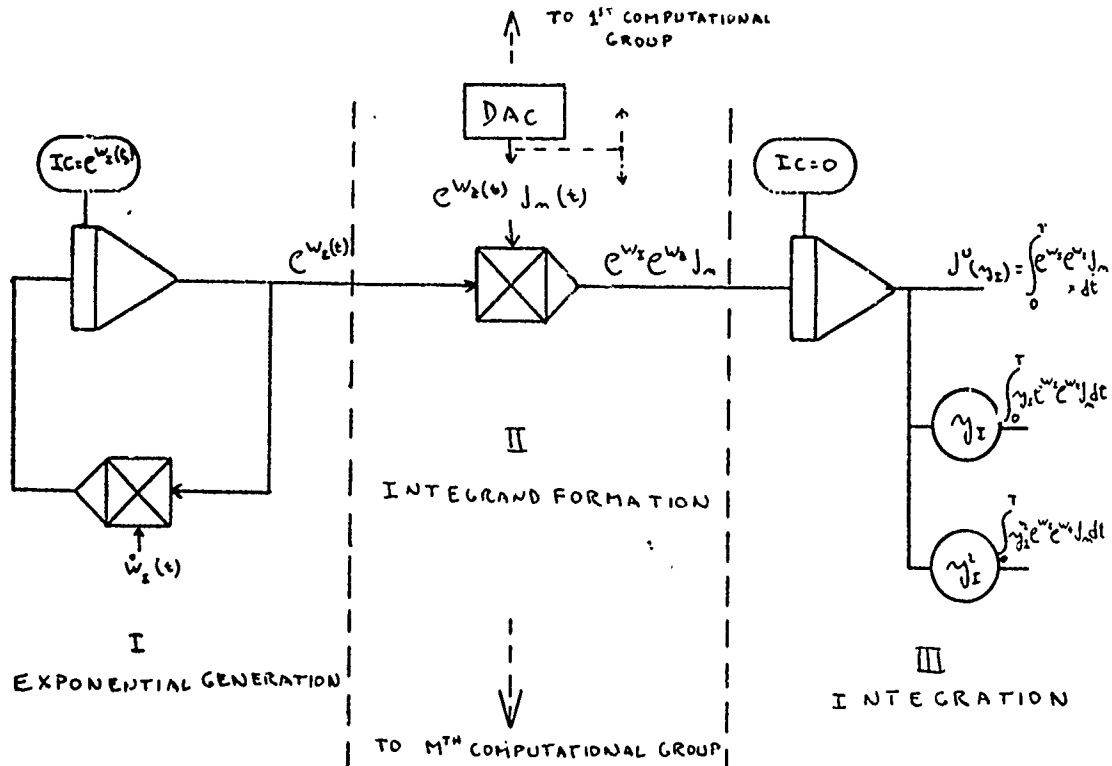


Figure 6. The Three Analog Computer Tasks In a Hybrid Optimal Filter: Exponential Generation (Section I), Integrand Formation (Section II), and Integration (Section III).

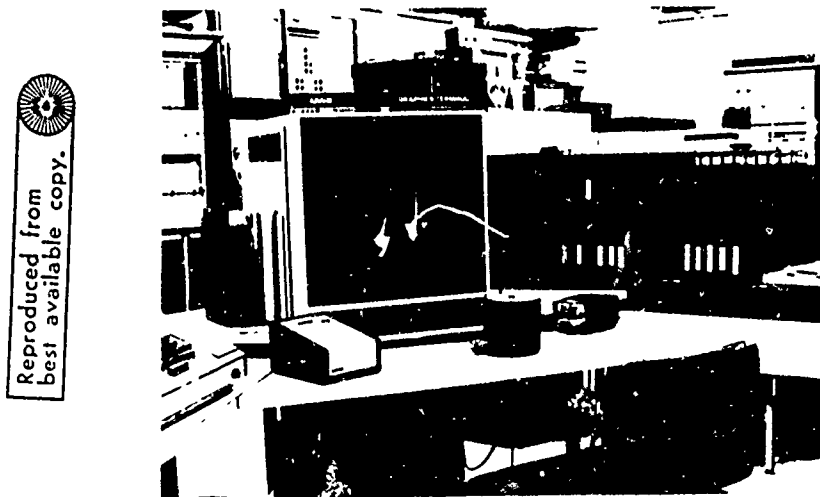


Figure 15. Adage Graphic Terminal, IBM 360-44 and a Density Function under Consideration as a Prior Distribution.

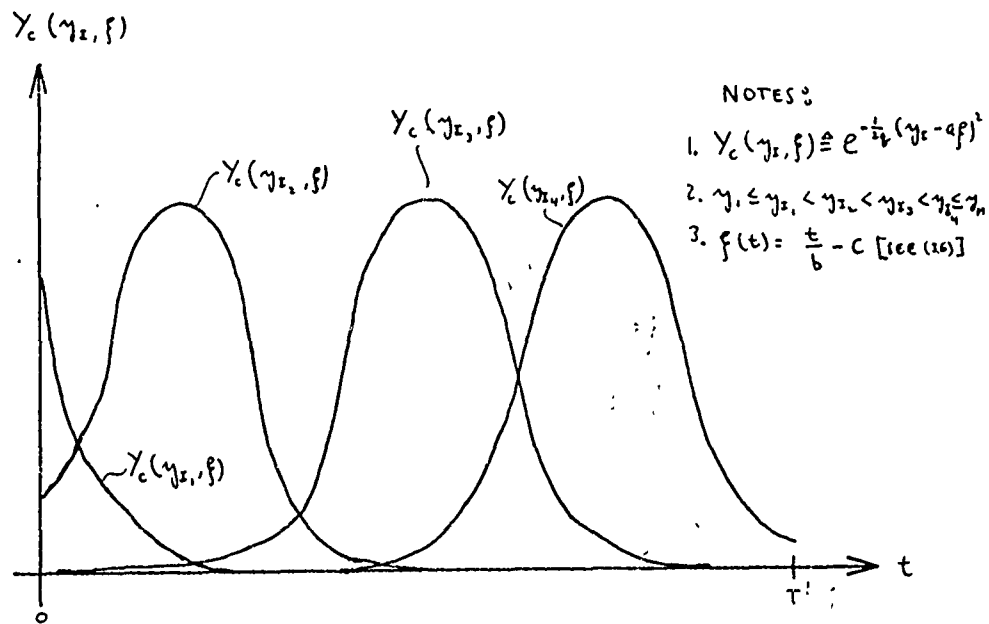


Figure 7. Exponentials Generated on an Analog Computer Corresponding to Four Typical $Y_c(\gamma_i, f)$ Terms, each of which Represents the Contribution of Model State Update Equation to the i^{th} Point of the $(n+1)^{\text{st}}$ Conditional Density Function.

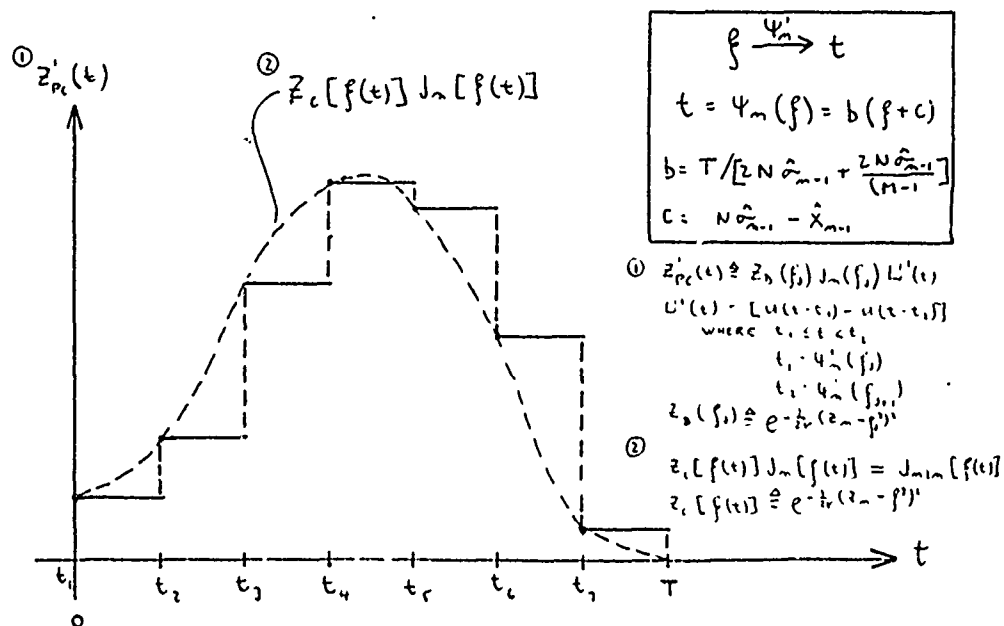


Figure 8. Piecewise Continuous Function $Z'_n(t)$ Appearing at Analog Computer as a Result of Digital to Analog Conversion in Accordance with the Map Ψ_n of Equation (3.4) for the Case $M=7$. Dotted Curve Represents a Mapping by Ψ_n of the Function $Z_c(f)J_n(f)$ which Coincides with $Z_n(t)$ at the M Computed Points.

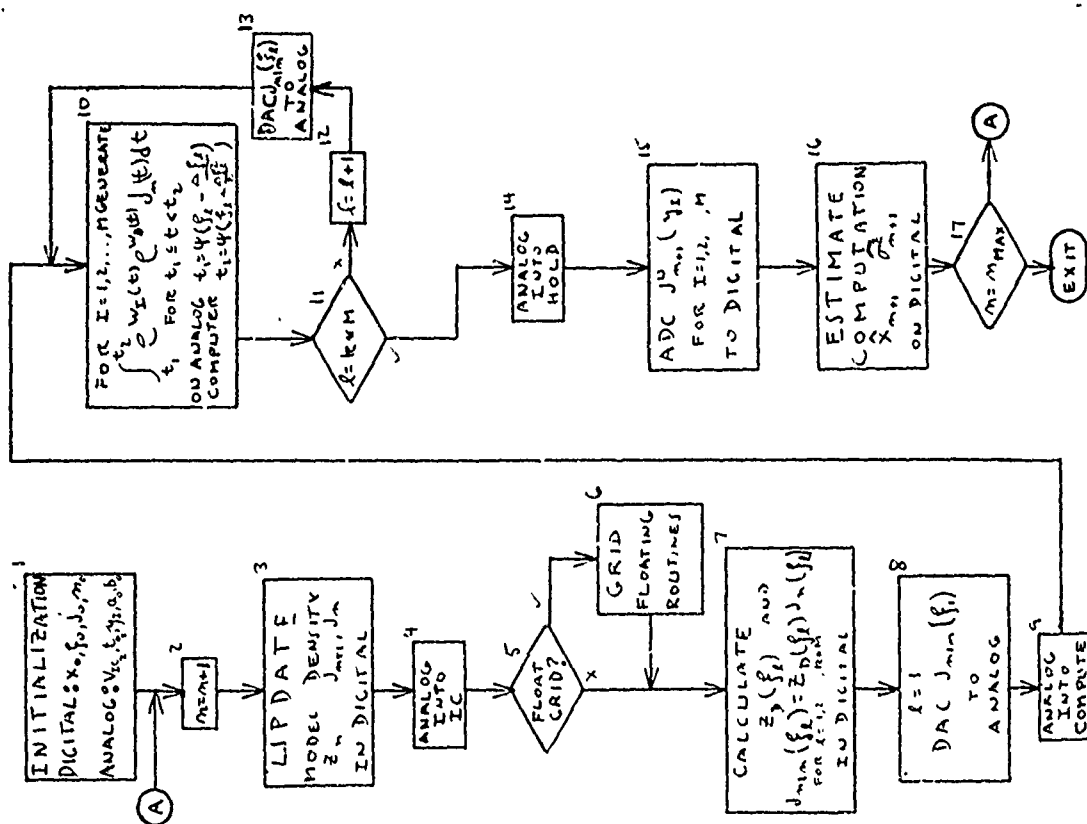


Figure 10. Flow Chart of the Hybrid Algorithm for Realization of an Optimal Discrete Nonlinear Filter as Applied to the Cubed Sensor Problem.

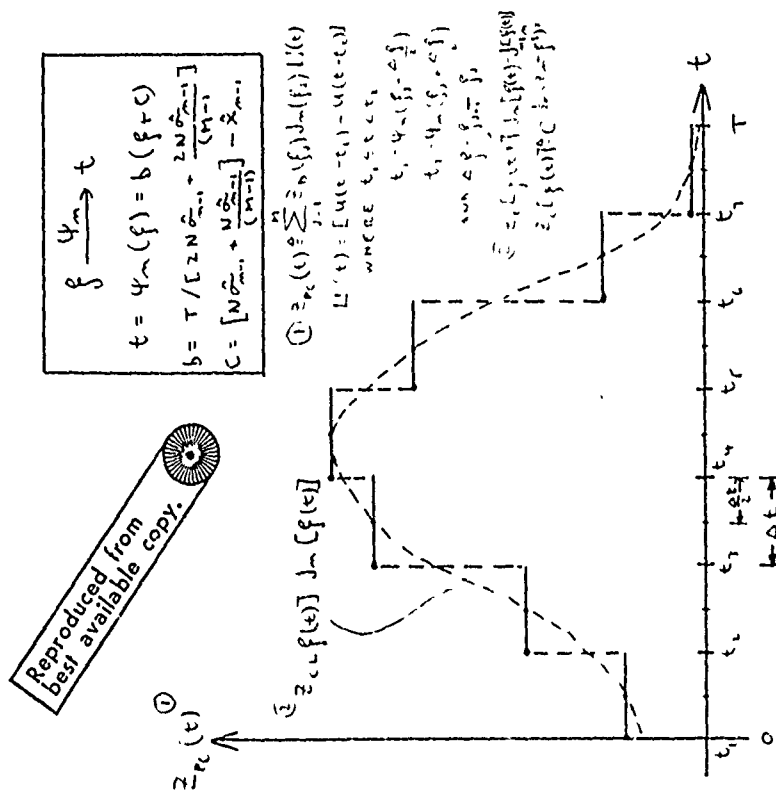


Figure 9. Piecewise Continuous Function $Z_k(t)$ Appearing at Analog Computer as a Result of Analog to Digital Conversion in Accordance with the Map Ψ_m of Equation (3.6) for the Case $M=7$. Dotted Curve Represents a Mapping by Ψ_m of the Function $Z_k(f) j_m(f)$ which Coincides with $Z_k(t)$ at the M Computed Points.

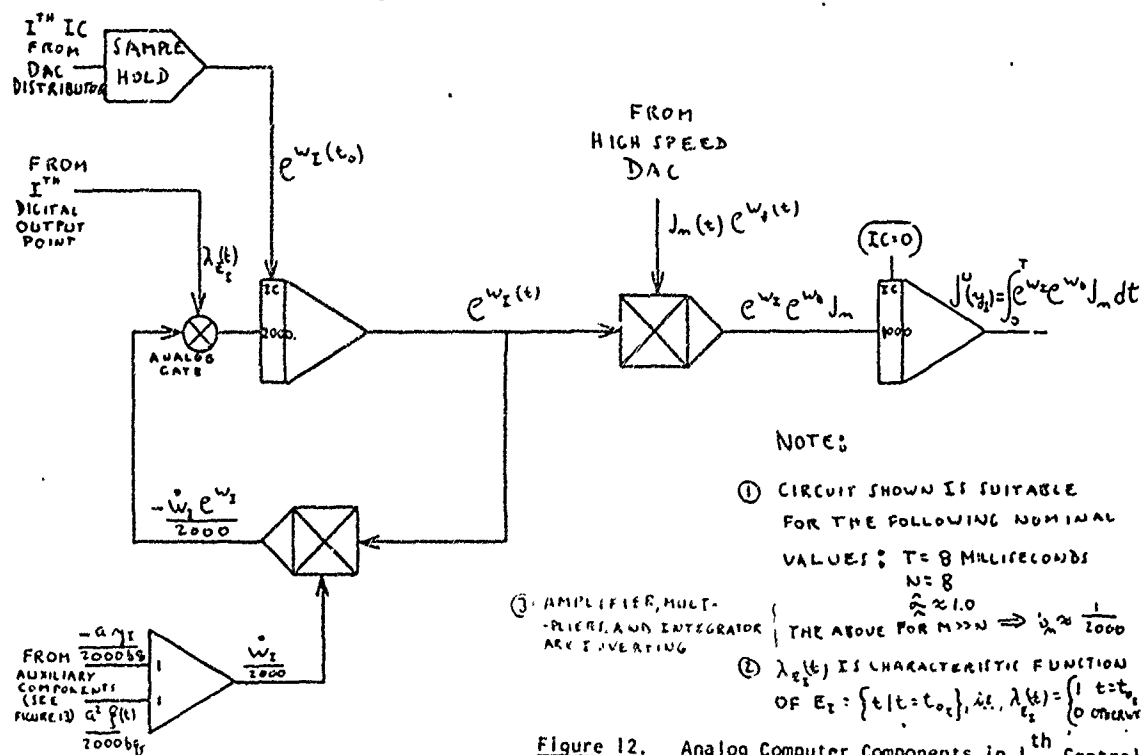
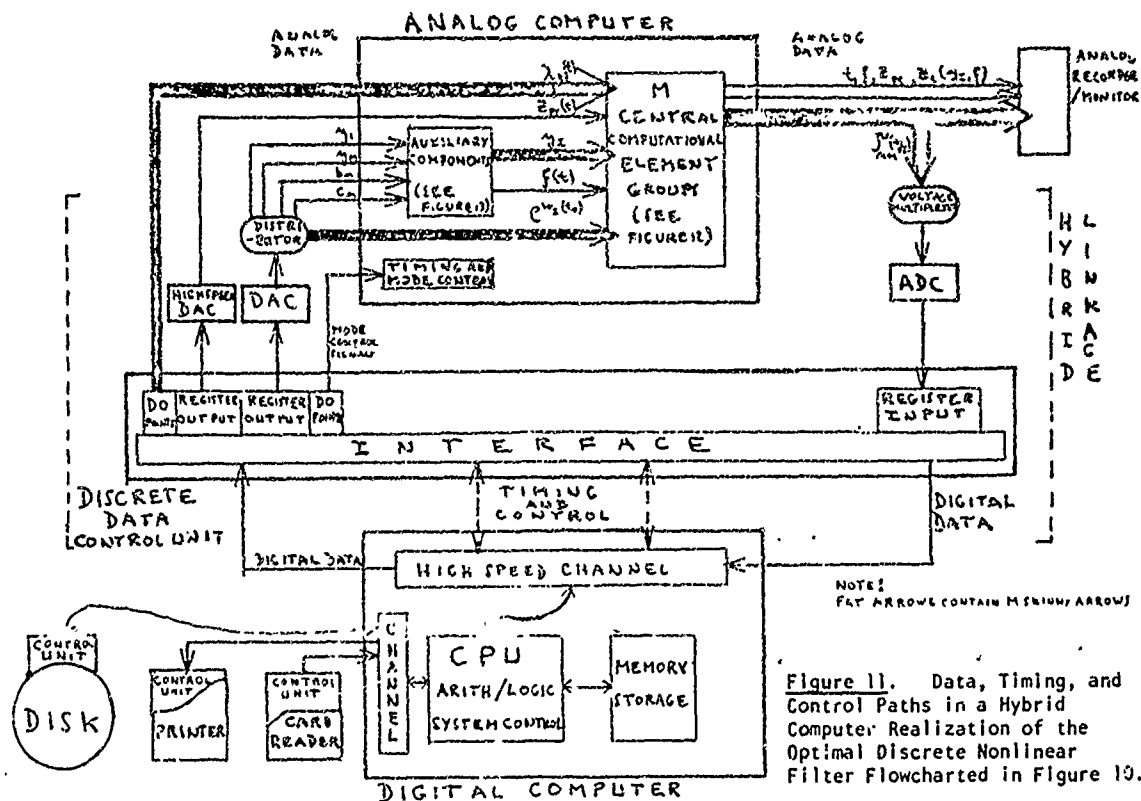


Figure 12. Analog Computer Components in 1st Central Computational Element Group of a Hybrid Optimal Discrete Filter for the Cubed Sensor Problem.

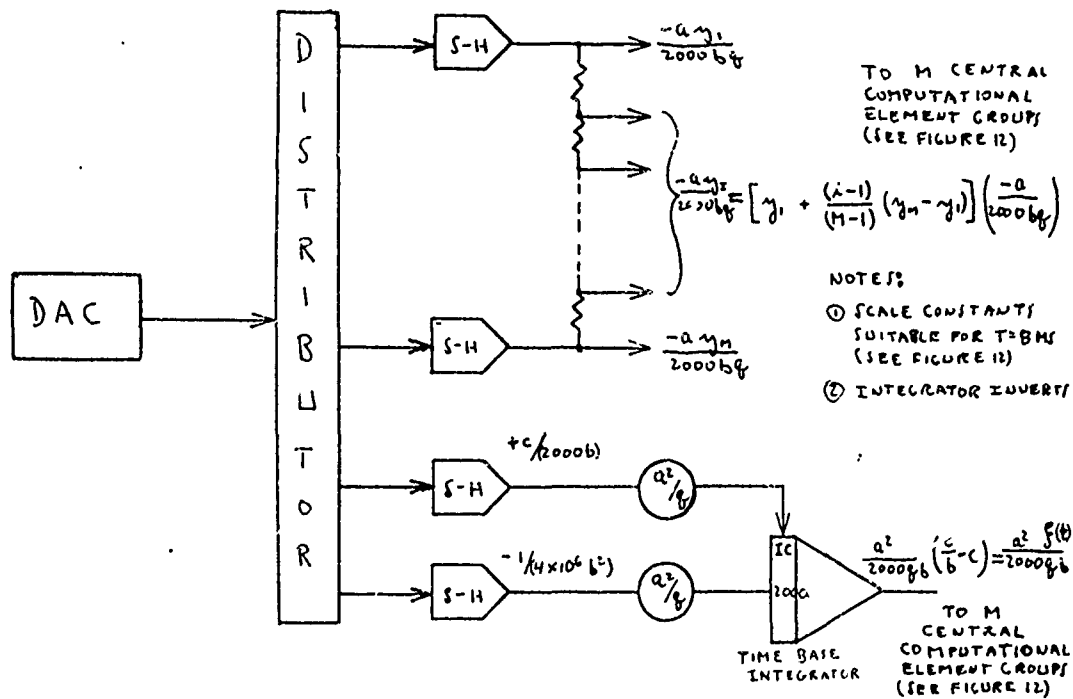


Figure 13. Analog Computer Components which Provide $\dot{w}_2(t)$ Terms for Feedback Multiplier in 1th Central Computational Element Group for Cubed Sensor Problem Shown in Figure 12.

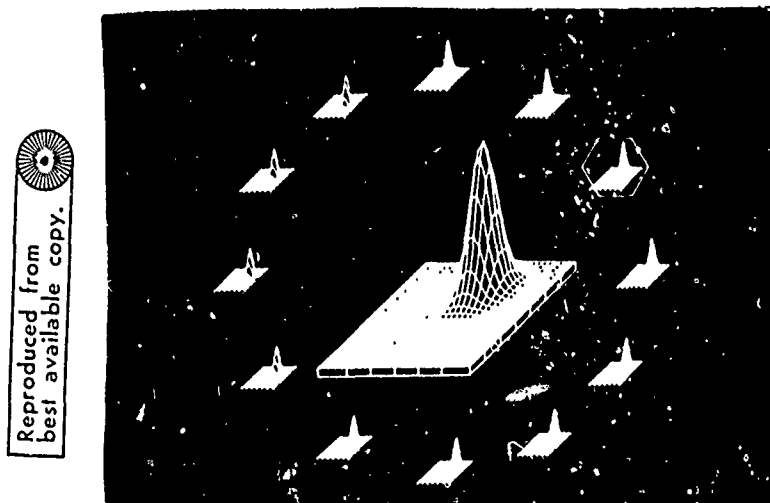


Figure 14. Multiple 2-D Density Function Display. Density function encircled by the hexagon is enlarged in the center of the screen.

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DIGITAL REALIZATION OF NON-LINEAR FILTERS*

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Abstract

A method is presented for solving the discrete-time non-linear estimation problem using Gauss-Hermite numerical integration and Hermite expansions of the conditional density function. Numerical results are given to demonstrate the effectiveness of the method.

1. INTRODUCTION

Equations have been developed which give recursive relations for the density function of the state vector conditioned on the observations (Bucy⁽²⁾). The equations are general and allow non-linear plant and observation processes, and non-Gaussian plant and observation noise terms. The equations involve integration over d -dimension space, and, except for $d = 1$, and in some cases 2, the equations have had no practical solutions.

Various approaches involving approximation techniques and the treatment of problems less general than the complete non-linear problem have been proposed. These approaches include schemes for approximating the non-Gaussian density function, methods of linearizing the system and treatment of problems with a non-changing state vector. Although the methods have had different levels of success, a good generalized method has not been produced.

In this paper a 2-dimensional problem is solved using a technique which appears capable of application to a six-dimension problem. The capability of the method depends on the extent of the non-linearities. The technique can be used for severe non-linearities, but the required number of points for numerical integration and the number of terms for the expansion of the density function become large, restricting the use for higher dimension problems.

2. BAYES LAW COMPUTATION

The equations for the conditional density of the state vector conditioned on the observation process using Bayesian estimation were derived in Bucy⁽⁴⁾. The relationships were for discrete time processes. The equations are also given in a slightly different form (avoidance of the pseudo-inverse) in Hecht⁽⁶⁾. The model, notation, and equations as given in the second reference are given in Tables 2.1 and 2.2.

The relations as given in Table 2.2 are generally very difficult to solve, either in closed form or

numerically. Most of the recent effort in non-linear estimation has been toward solving the recurrence equations for specific applications. Several papers at the Symposium on Non-Linear Estimation Theory of 1970⁽¹²⁾ dealt with numerical techniques or other schemes to obtain practical solutions. It can be seen from equation (2.4), for example, that in order to construct a density function $J_n/n(y)$ defined on a set of points y_i , it would be required to integrate over the entire d -dimension space R^d . If the prior density function were defined on a domain which consisted of k -points for each component of x , it would be equal to k^d and the number of computations for one update of the density function would be $(k^d)^2$. For k and d larger than approximately 20 and 2, respectively, the computation would be impossible with contemporary computers.

Bucy and Senne⁽³⁾ have devised a method of defining the function and performing the integration by selecting the domain in a special way. By the efficient selection of the points on which to define the density function they were able to solve a two-dimensional problem using only $(15)^2$ points and achieved an accuracy equivalent to approximately $(56)^2$ points, with a corresponding decrease in computation time. Bucy, Moller and Merritt⁽⁵⁾ have proposed reducing computation time by performing the integrations in parallel on an analog computer as part of a hybrid computation arrangement. Lo⁽⁹⁾ and Alspach and Sorenson⁽¹⁾ represent the density by a sum of Gaussian functions as another approach, although the true non-linear estimation is not achieved in the latter paper because only a linearized system is considered. Kizner⁽⁸⁾ and others represent the density function with Hermite function expansions to solve the non-linear problem, but also fail to demonstrate the Bayes rule recurrence relation with a non-linear system and plant driving terms.

In the following development, several techniques are employed for practical solutions to the equations of Table 2.2. The next section gives the

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analytic details of a two-dimension example where the observation process, $h_n(x)$, is non-linear and the plant is linear. Both the plant and observation noise are Gaussian. By making a Hermite expansion of the non-linear prior density function, the a posteriori function is computed in closed-form. To obtain the Hermite coefficients for the new density function requires numerical integration, but the Gauss-Hermite numerical integration formulae are utilized. The Gauss-Hermite formula has been demonstrated in Hecht to be eight times more effective than rectangular

integration for a one-dimension example. That is, equivalent accuracy was achieved using one-eighth the number of points to define the function to be integrated.

The technique for the two-dimension case can be extended to higher dimension examples. In the higher dimension case, computation time is reduced by performing a portion of the integrations analytically, with a potential of reducing the computation time for a d-dimension problem to the equivalent of a d/2-dimension problem.

Table 2.1 - Model for Bayes Rule Conditional Density Recursion Formula

$$x_n = \Phi_n(x_{n-1}) + \sigma_{n-1}(x_{n-1}) u_{n-1} \quad (2.1)$$

$$x_0 = c \quad (2.2)$$

$$z_n = h_n(x_n) + v_n \quad (2.3)$$

Notation

x_n	= a sequence of d-dimension random vectors
n	= time index
$\Phi_n(x_{n-1})$	= a function from R^d to R^d
$\sigma_{n-1}(x_{n-1})$	= a function from R^d to $d \times r$ matrices
$\{u_n\}$	= a process of independent r-dimension random vectors with density $p_{u_n}(\omega)$.
c	= a d-dimension random vector independent of the u_n process and having density $p_c(X)$.
z_n	= a sequence of s-dimension random vectors
$h_n(x_n)$	= a function from R^d to R^s
$\{v_n\}$	= a process of independent s-dimension random vectors with density $p_{v_n}(\phi)$, independent of c and the u_n process.

Table 2.2 - Conditional Density Recursion Formulae for Bayesian Estimation

Recurrence Relation	Equation for Conditional Density
$J_{n-1/n-1}(y) \rightarrow J_{n/n}(y)$	$J_{n/n}(y) = \frac{1}{k(n)} p_{v_n}(z_n - h(y)) \int \cdot d \cdot \int p_{\sigma_u}(y - \Phi(x)) J_{n-1/n-1}(x) dx \quad (2.4)$
$J_{n/n-1}(y) \rightarrow J_{n/n}(y)$	$J_{n/n}(y) = \frac{1}{k(n)} p_{v_n}(z_n - h(y)) J_{n/n-1}(y) \quad (2.5)$
$J_{n/n-1}(y) \rightarrow J_{n+1/n}(y)$	$J_{n+1/n}(y) = \frac{1}{k(n)} \int \cdot d \cdot \int p_{\sigma_u}(y - \Phi(x)) p_{v_n}(z_n - h(x)) J_{n/n-1}(x) dx \quad (2.6)$
$J_{n/n}(y) \rightarrow J_{n+1/n}(y)$	$J_{n+1/n}(y) = \frac{1}{k(n)} \int \cdot d \cdot \int p_{\sigma_u}(y - \Phi(x)) J_{n/n}(x) dx \quad (2.7)$

Note: The index n has been omitted in some places for notational simplicity.

Notation:

$J_{a/b}(y)$	= conditional density of the random vector x_a given observation vectors $z_0 \dots z_b$
$k(n)$	= a normalizing constant to make $J(\cdot)$ a density function
$\int \cdot d \cdot \int (\cdot) dx$	= d-fold integration with respect to the variables x_i , components of the vector x .
$p_{\sigma_u}(\cdot)$	= density of the d-dimension random vector $\sigma_{n-1}(x_{n-1}) u_{n-1}$

3. TWO-DIMENSIONAL REALIZATION

A non-linear filter was synthesized to solve a phase angle receiver/demodulator problem. The model for the system consisted of position and velocity. The receiver, described in Viterbi (13), and Mallinckrodt, Bucy, Cheng (10), was modeled as observing the cosine and sine of the position coordinate.

The equations to describe the signal process, or plant equations, were

$$\begin{cases} \dot{x}_1 = x_2 + \varepsilon \omega_1 \\ \dot{x}_2 = \omega_2 \end{cases} \quad (3.1)$$

where $\varepsilon \omega_1$ and ω_2 were uncorrelated Gaussian white noise processes. These equations were converted to discrete time and took the form

$$\begin{cases} x_1(n) = x_1(n-1) + x_2(n-1)\Delta + \varepsilon u_1(n-1)\Delta \\ x_2(n) = x_2(n-1) + u_2(n-1)\Delta \end{cases} \quad (3.2)$$

Hecht (6) gives a summary of modeling with discrete time systems and the conversions between discrete and continuous time models. Δ is the time between samples.

Rewriting equation (3.2),

$$x(n+1) = Fx(n) + Gu(n) \quad (3.3)$$

$$\text{with } F = \begin{bmatrix} 1 & \Delta \\ 0 & 1 \end{bmatrix} \quad G = \begin{bmatrix} \varepsilon & 0 \\ 0 & 1 \end{bmatrix}$$

$$x = \begin{bmatrix} x_1(n) \\ x_2(n) \end{bmatrix} \quad u = \begin{bmatrix} u_1(n) \\ u_2(n) \end{bmatrix}$$

The observation process was

$$z(n) = H(x_1(n), x_2(n)) + v(n) \quad (3.4)$$

with

$$H = \begin{bmatrix} \cos x_1(n) \\ \sin x_1(n) \end{bmatrix}$$

$$z(n) = \begin{bmatrix} z_1(n) \\ z_2(n) \end{bmatrix} \quad v(n) = \begin{bmatrix} v_1(n) \\ v_2(n) \end{bmatrix}$$

The density functions, using the notation of Table 2.2, were

$$p_{Gu}(\varepsilon) = \frac{1}{(2\pi)^{1/2} |Q|^{1/2}} \exp \left\{ -\frac{1}{2} \|\varepsilon\|_{Q^{-1}}^2 \right\} \quad (2.5)$$

$$p_v(\varepsilon) = \frac{1}{(2\pi)^{1/2} |R|^{1/2}} \exp \left\{ -\frac{1}{2} \|\varepsilon\|_{R^{-1}}^2 \right\} \quad (3.6)$$

$$\text{where } Q = \begin{bmatrix} E(\varepsilon^2) & 0 \\ 0 & E(u_2^2) \end{bmatrix} = \begin{bmatrix} \varepsilon^2 q & 0 \\ 0 & q \end{bmatrix}$$

$$R = \begin{bmatrix} E[v_1^2] & 0 \\ 0 & E[v_2^2] \end{bmatrix} = \begin{bmatrix} r & 0 \\ 0 & r \end{bmatrix}$$

In the signal process, the position is a deterministic function of velocity, which is accounted for by letting $\varepsilon \rightarrow 0$.

The filter problem was solved by applying equa-

tions (2.5) and (2.7) of Table 2.2:

$$J_{n+1/n}(y) = \frac{1}{K(n)} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} p_{Gu}(y-Fx) J_{n/n}(x) dx_1 dx_2 \quad (3.7)$$

$$J_{n/n}(y) = \frac{1}{K(n)} p_v(z_n - H(y)) J_{n/n-1}(y) \quad (3.8)$$

In the expression $p_{Gu}(\cdot)$ of equation (3.7) let $\varepsilon \rightarrow 0$ to obtain

$$\lim_{\varepsilon \rightarrow 0} p_{Gu}(y-Fx) = c_0 \delta(y_1 - x_2 \Delta - x_1) \exp \left\{ -\frac{1}{2q} (y_2 - x_2)^2 \right\} \quad (3.9)$$

where $c_0 = \frac{1}{(2\pi)^{1/2} q^{1/2}}$

$\delta(\cdot)$ = dirac delta function.

Equation (3.9) was put into (3.7).

$$J_{n+1/n}(y) = c_1 \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \delta(y_1 - x_2 \Delta - x_1) \exp \left\{ -\frac{1}{2q} (y_2 - x_2)^2 \right\} J_{n/n}(x) dx_1 dx_2$$

with c_1 a generalized constant which includes $k(n)$ and c_0 ; in general, we let c_1 be a generalized constant which may include contributions from several sources.

The integration with respect to x_1 was performed by taking advantage of the δ -function to obtain

$$J_{n+1/n}(y) = c_1 \int_{-\infty}^{\infty} \exp \left\{ -\frac{1}{2q} (y_2 - x_2)^2 \right\} J_{n/n} \left(\begin{matrix} y_1 - x_2 \Delta \\ x_2 \end{matrix} \right) dx_2 \quad (3.10)$$

$$J_{n/n} \left(\begin{matrix} y_1 \\ y_2 \end{matrix} \right) = c_2 \exp \left\{ -\frac{1}{2r} [(z_1 - \cos y_1)^2 + (z_2 - \sin y_1)^2] \right\} J_{n/n-1} \left(\begin{matrix} y_1 \\ y_2 \end{matrix} \right) \quad (3.11)$$

The pair of equations, (3.10) and (3.11) give the required conditional density filter update relations. The original two-dimension problem, involving the estimation of a 2-vector, was reduced to a single integration due to the fact that only one of the states was driven by a random process.

In general, when the signal process is a dynamic system with the elements of the state vector representing position and velocity, the covariance matrix will be singular since the only driving terms are acceleration. Thus to solve a problem in three-dimension physical space requires a six-dimension state vector, but the problem can be reduced to a triple integration in the same manner as above. Bucy and Senne (3) mention the computational simplification resulting from the singular covariance matrix. It is the above simplification to which they are referring.

The integration of equation (3.10) is by no means an easy task. Normally, when numerical integration is applied, a posteriori density function

$J_{n/n} \left(\begin{matrix} y_1 \\ y_2 \end{matrix} \right)$ is computed on a set of points $y_1(i)$, $y_2(i)$. This function becomes the prior density function for the subsequent iteration. What is needed for equation (3.10), however, is

$J_{n/n} \begin{bmatrix} y_1 - x_2 \\ x_2 \end{bmatrix}$. That is, the function is needed with the first argument different than the one for which the function was computed.

By use of Hermite polynomial expansions the above difficulty is surmounted and an explicit formula was derived for the new density function which involves no integrations.

Because of space limitations in this paper the detailed steps of the development are not given. An outline of the steps follows with equation (3.12) the final result. We assume a set of quantities are available at a particular time which completely characterizes the conditional density function, and demonstrate how these same quantities are computed for the following time increment.

1. From the prior stage we have the means, covariance matrix, characteristic values and vectors of the covariance matrix, and a series expansion of the conditional density in characteristic vector coordinates in terms of Hermite polynomials.
2. Series form of the density is put into equation (3.10) and then one takes the Fourier transform of both sides.
3. Rework the result of Step 2 to take the inverse Fourier transform to obtain $J_{n+1/n}$.
4. Multiply $J_{n/n-1}$ by the observation process, equation (3.11), to obtain $J_{n/n}$.
5. Compute moments for means, covariance matrix, and for the coefficients of the new expansion.

$$J_{n/n} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = c_3 \cdot \exp \left\{ -\frac{1}{2r} \left[(z_1 - \cos y_1)^2 + (z_2 - \sin y_1)^2 \right] \right\} \cdot \exp \left\{ -\frac{1}{2A} \left(y_2 - (T_2 y_1 - T_1 \bar{v}_1) \right)^2 \right\} \cdot \exp \left\{ -\frac{1}{2} \frac{B^2}{c^2} \left[y_2 - \beta(T_2 y_1 - T_1 \bar{v}_1) + \frac{1}{B} (T_4 y_1 - T_3 \bar{v}_2) \right]^2 \right\} \cdot \sum_{r=0}^{m_1} \sum_{\ell=0}^{m_2} b_{r\ell} (\text{sign } T_1)^r (\text{sign } T_2)^\ell (B)^{\frac{r+\ell}{2}} \left(\frac{\lambda_2 T_3^2 A}{\lambda_2 T_3^2 q + \lambda_1 \lambda_2 T_1^2 T_2^2 + \lambda_1 T_1^2 q} \right)^{\frac{\ell+1}{2}} \cdot \sum_{k=0}^r \binom{r}{k} (-1)^k \left[\frac{q^2}{\lambda_2 T_3^2 q + \lambda_1 \lambda_2 T_1^2 T_2^2 + \lambda_1 T_1^2 q} \right]^{k/2} H_{r-k} \left(\frac{1}{\sqrt{2A}} \left(y_2 - (T_2 y_1 - T_1 \bar{v}_1) \right) \right) \cdot H_{k+\ell} \left(\frac{B}{\sqrt{2c}} \left(y_2 - \beta(T_2 y_1 - T_1 \bar{v}_1) + \frac{1}{B} (T_4 y_1 - T_3 \bar{v}_2) \right) \right). \quad (3.12)$$

where all of the undefined terms are functions of the quantities listed in Step 1, and c_3 is a normalizing constant to make $J_{n/n}$ a density function.

$J_{n/n}$, as given by equation (3.12) is the required equation to update the conditional density function for each sampling time. To complete the cycle it is necessary to approximate the density function with a new Hermite series, as characterized by a new set of coefficients, $b_{r\ell}$. This operation is shown in the next section.

4. HERMITE APPROXIMATION AND NUMERICAL INTEGRATION

Under the proper conditions a function of two variables, $f(x_1, x_2)$ can be approximated by a series expansion of Hermite polynomials of the form

$$f(x_1, x_2) \approx y(x_1, x_2) = e^{-\alpha_1^2 x_1^2} e^{-\alpha_2^2 x_2^2} \sum_{r=0}^{m_1} \sum_{\ell=0}^{m_2} b_{r\ell} H_r(\alpha_1 x_1) \cdot H_\ell(\alpha_2 x_2) \quad (4.1)$$

with the coefficients, $b_{r\ell}$, determined by

$$b_{r\ell} = \frac{\alpha_1 \alpha_2}{2^r r! 2^\ell \ell! \pi} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x_1, x_2) H_r(\alpha_1 x_1) H_\ell(\alpha_2 x_2) dx_1 dx_2 \quad (4.2)$$

The approximation is the best in the sense that

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-\alpha_1^2 x_1^2} e^{-\alpha_2^2 x_2^2} \left[f(x_1, x_2) - y(x_1, x_2) \right]^2 dx_1 dx_2 = \text{minimum} \quad (4.3)$$

By letting $f(x_1, x_2)$ be the density function of the previous section with

$$\alpha_1^2 = \frac{1}{2\sigma_1^2}, \quad \alpha_2^2 = \frac{1}{2\sigma_2^2}, \quad \sigma_1^2, \sigma_2^2$$

the characteristic values of the covariance matrix, the conditions for the approximation to be valid are easily satisfied (Hildebrand (7)).

Let $f(x_1, x_2)$ be a density function we wish to approximate, and let $m_1 = m_2 = 0$ in equation (4.1). Then

$$b_{00} = \frac{\alpha_1 \alpha_2}{\pi} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x_1, x_2) dx_1 dx_2 = \frac{\alpha_1 \alpha_2}{\pi} \quad (4.4)$$

$$y(x_1, x_2) = \frac{\alpha_1 \alpha_2}{\pi} e^{-\alpha_1^2 x_1^2} e^{-\alpha_2^2 x_2^2} \quad (4.5)$$

Let σ_1^2 and σ_2^2 be the variances of $f(x_1, x_2)$, and let

$$\alpha_1^2 = \frac{1}{2\sigma_1^2}, \quad \alpha_2^2 = \frac{1}{2\sigma_2^2}$$

Equation (4.5) is

$$y(x_1, x_2) = \frac{1}{2\pi\sqrt{\sigma_1\sigma_2}} e^{-\frac{x_1^2}{2\sigma_1^2} - \frac{x_2^2}{2\sigma_2^2}} \quad (4.6)$$

By taking only the zero'th term of the expansion we can get a Gaussian fit to the true density function.

In addition to the above assumptions, except let m_1 and $m_2 > 0$, let η_1 and η_2 be the means of $f(x_1, x_2)$ and μ_{ij} the i - j 'th central moment.

$$\mu_{ij} = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x_1 - \eta_1)^i (x_2 - \eta_2)^j f(x_1, x_2) dx_1 dx_2 \quad (4.7)$$

The product of the Hermite polynomials when expanding about the expected values, can be written

$$H_r(\alpha_1(x_1 - \eta_1)) \cdot H_\ell(\alpha_2(x_2 - \eta_2)) = \sum_{i=0}^r \sum_{j=0}^{\ell} a_i(r) a_j(\ell) \alpha_1^i \alpha_2^j (x_1 - \eta_1)^i (x_2 - \eta_2)^j \quad (4.8)$$

Equation (4.2) is then

$$b_{r\ell} = \frac{\alpha_1^r \alpha_2^\ell}{2^r r! 2^\ell \ell! \pi} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x_1, x_2) \sum_{i=0}^r \sum_{j=0}^{\ell} a_i(r) a_j(\ell) \alpha_1^i \alpha_2^j (x_1 - \eta_1)^i (x_2 - \eta_2)^j dx_1 dx_2$$

$$= \frac{\alpha_1^r \alpha_2^\ell}{2^r r! 2^\ell \ell! \pi} \sum_{i=0}^r \sum_{j=0}^{\ell} a_i(r) a_j(\ell) \alpha_1^i \alpha_2^j \mu_{ij} \quad (4.9)$$

The coefficients $a_i(r)$ $a_j(\ell)$ in (4.9) are functions of the Hermite polynomial coefficients, and need to be computed only one time in advance and stored.

The recurrence formula for the Hermite polynomials is (Hildebrand (7)):

$$\left. \begin{aligned} H_{r+1}(x) &= 2x H_r(x) - 2r H_{r-1}(x) \\ H_0(x) &= 1 \quad H_1(x) = 2x \end{aligned} \right\} \quad (4.10)$$

Designating c_{mn} as the coefficient of x^n in $H_m(x)$, $a_i(r)$ $a_j(\ell)$ in equation (4.9) is determined from

$$a_i(r) a_j(\ell) = c_{ri} c_{\ell j} \quad (4.11)$$

c_{mn} was generated from a recurrence formula which was derived from (4.10).

Comparing equation (4.1) with the conditional density function in eigenvector coordinates from the previous section, we see the approximating polynomial expansion can be accomplished by letting:

$$\alpha_1 = \frac{1}{2\lambda_1} \quad \alpha_2 = \frac{1}{2\lambda_2}$$

$$\eta_1 = \bar{v}_1 \quad \eta_2 = \bar{v}_2$$

and using (4.9) to compute the coefficients $b_{r\ell}$.

Gauss-Hermite numerical integration was used to perform the integrations for the moments of equation (4.7). The conditional density function was given by equation (3.12).

Gauss-Hermite integration is performed according to the following formula:

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-x_1^2 - x_2^2} f(x_1, x_2) dx_1 dx_2$$

$$\approx \sum_{i=1}^n \sum_{j=1}^n \omega_i \omega_j f(x_{1i}, x_{2j}) \quad (4.12)$$

with x_{1i} and x_{2j} the roots of $H_n(x)$ and ω_i and ω_j the corresponding weighting functions. The numbers ω_i and x_i are tabulated in several places for a variety of values of n ; Shao, Chen, Frank (11) for example.

The a posteriori density function of equation (3.12) consists of a product of an exponential function and a polynomial function. The exponential function has two contributions, the non-linear observation part with the trigonometrical functions and a Gaussian part.

To perform the numerical integration of equation (3.12) effectively, it was found desirable to consider two situations and to treat them differently.

1) When the contribution of the observation process was small (r large), the non-Gaussian exponential was combined with the polynomial function to construct $f(x_1, x_2)$ of equation (4.12). The characteristic values and vectors of the remaining Gaussian part were found and used to formulate the exponential part of the integrand of (4.12). 2) When the contribution of the observation process was significant (r small), the non-Gaussian exponent was linearized. The linearized exponent was combined with the other Gaussian exponents and treated in the same way as in 1). The difference between the non-linear and linear exponent was combined with the polynomial function to construct $f(x_1, x_2)$ of equation (4.12).

5. NUMERICAL RESULTS

The equations of the previous sections were programmed and a limited number of runs were made. An extended Kalman-Bucy filter (extended K-B filter) was programmed for the same problem using the same noise sequence for comparison purposes.

Runs were made using the non-linear filter with the coefficients, $b_{r\ell}$ of equation (4.12), up to b_{88} . It was found that equal accuracy could be obtained by eliminating higher order cross-

coefficients, so by adjusting m_1 and m_2 in equation (3.12) so that $r+l \leq 8$, a reduction in running time without loss of accuracy was obtained.

The numerical integration was performed using 10-point and 20-point integrations. For good signal-to-noise ratios, the lower number of points was adequate, whereas with poor ratios more accurate integration was needed. This was to be expected since for good signal/noise the equations are close to being Gaussian and the Gauss-Hermite integration formulae are almost perfect. The basic independent parameter for the numerical trials was the continuous linear model error variance of x_1 , designated $p_{11}(o)$. The discrete linearized variance of the one-step prediction is designated $p_{11}(\Delta)$, and in the limit $p_{11}(\Delta) \rightarrow p_{11}(o)$. In Hecht⁽⁶⁾ it is shown that for Δ small compared to the filter time constant, F ,

$$p_{11}(\Delta) \approx p_{11}(o) \left(1 + \frac{\Delta}{F}\right).$$

All trials were made with $\Delta/F = 0.1$.

Viterbi⁽¹³⁾ shows $p_{11}(o)$ is the inverse of the effective signal/noise ratio, so $p_{11}(o)$ has the double meaning for the continuous linearized model:

$$1) p_{11}(o) = E \left[(x_1 - \hat{x}_1)^2 \right]$$

$$2) p_{11}(o) = \frac{1}{S/N} = \frac{N}{S}$$

Trials were made with sample path of 100 points (10 filter time constants). The variance of the initial state was set equal to r in every case. The signal term, q , was set to 0.01 for all trials.

For very small $p_{11}(o)$ the extended K-B filter and the non-linear filter performed almost identically. For $p_{11}(o) = .0001$ the sequence of estimates for the same signal and noise sequences agreed to within 6 decimal places.

As the parameter $p_{11}(o)$ was increased, the non-linear filter was also tested using only one coefficient, b_{00} . The reason for this is that one would expect better performance from the one-coefficient filter than the extended K-B filter due to the integration of the non-Gaussian term. It was desired to determine the performance of the nine-coefficient filter since all of the non-linear information is carried by the coefficients not including b_{00} . By checking both the one and nine-coefficient filter, the additional improvement of the higher terms was evaluated.

For $p_{11}(o) \leq .04$ ($S/N = 13$ db) the average variance for a 100 sample run was almost the same for the extended K-B, the one-, and the nine-coefficient filters. For $p_{11}(o) = .16$ ($S/N = 8$ db), the one-coefficient filter shows an improvement over the extended K-B, and the nine-coefficient filter shows an equal improvement over the one-coefficient filter.

For $p_{11}(o) = .25$ ($S/N = 6$ db), the non-linear filter appears substantially better than the extended K-B. These results, although for a very small sampling, do indicate an encouraging trend. They are summarized in Table 5.1.

Table 5.1 - Average Variance for Extended Kalman-Bucy, One-Coefficient, and Nine-Coefficient Non-Linear Filters.

Sample Path = 100 points

$p_{11}(o)$	Q_D	R_D	Δ	K-B	1-Coef.	9-Coef.
.01	.00126	0.05	0.126	6.6852 E-3	7.0754 E-3	7.0782 E-3
.04	.00200	0.2	0.200	6.1703 E-2	6.1681 E-2	6.1711 E-2
.16	.00317	0.8	0.317	2.0403 E-1	1.9556 E-1	1.8349 E-1
.25	.00368	1.25	0.368	7.2118 E-1	2.9438 E-1	2.7058 E-1

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COMPUTER EXPERIMENTS
WITH NONLINEAR ESTIMATORS*

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Abstract

Elaborate and extensive experiments with various digital implementations of discrete-time nonlinear estimators reveals some interesting and important characteristics of nonlinear estimators. The experiments include extensive Monte Carlo comparisons of alternative techniques, sample path comparisons of optimal and approximate estimators and associated graphical illustration of relevant considerations and details.

1. INTRODUCTION

In the proceedings of the first Symposium on Nonlinear Estimation Theory, Bucy and Senne proposed a digital realization for nonlinear estimators (henceforth referred to as the Floating Grid Method)^{1,2}. The basis for the Floating Grid Method is a numerical approximation to the general Bayes-optimal solution to the nonlinear estimation problem³. Thus the Floating Grid is one of a variety of possible applications of numerical techniques to calculate the Bayes integral recursion formula^{4,5}. The alternatives to the numerical techniques, which involve only density representation and integral approximations, are the various analytical techniques, which involve functional approximations of the underlying physical problem equations and/or moment series expansions of the

conditional densities^{3,6,7}. Common analytical approximations include linearized (extended) Kalman-Bucy filters, quasi-moment filters, second order techniques, etc. The obvious advantages of analytical approximations to the estimation problem are the relative simplicity of the computations, compared with the numerical approximations. On the other hand, improvements to analytical techniques are relatively hard to obtain since generalizations come at the expense of extensive analysis and/or engineering experimentation, whereas the accuracy of a numerical approximation can be directly related to computation time, with little or no modification to the computational algorithm itself. Thus, while practical applications will frequently demand the simplicity of analytical approximations, the answers to questions relating to the behavior

* The views expressed herein are those of the author and do not necessarily reflect the views of the United States Air Force or the Department of Defense.

of optimal estimates which can give considerable insight into the selection of practical estimators may best be obtained by studying a suitable numerical approximation.

In this presentation a report is given describing extensive additional experimentation with the Floating Grid Method and comparisons with analytical approximates including the linearized filter and the generalized linearized filter of Alspach and Sorenson⁵. All techniques have been exposed to the same data records both for sample path studies and elaborate Monte Carlo performance comparisons. As a result of the simulations, many conclusions have been made concerning the behavior of optimal estimates, the "engineering modifications" to analytic approximations which improve performance, and experimental methods relating to evaluation of estimators. Sample path studies are exemplified by a movie illustrating the evolution of the conditional probability densities for a passive tracking problem. Monte Carlo studies are given to compare the performance of the Floating Grid Method, the linearized estimator, the "modified" linearized estimator, and the generalized linearized (gaussian-sum) estimator on a similar tracking problem. The results illustrate conclusively that additional efforts on sophisticated numerical techniques will result in considerable additional insight and experience with optimal nonlinear estimators.

2. PROBLEM DESCRIPTION

Consider the non-linear dynamical system described by

$$\begin{aligned} \underline{x}_n &= \underline{f}_n(\underline{x}_{n-1}) + \sigma_n(\underline{x}_{n-1}) \underline{u}_{n-1}, & \underline{x}_0 &= \underline{c} \\ \underline{z}_n &= \underline{h}_n(\underline{x}_n) + \underline{v}_n \end{aligned} \quad (1)$$

with

$$\begin{aligned} \underline{f}_n(\cdot) &: \mathbb{R}^d \rightarrow \mathbb{R}^d \\ \sigma_n(\cdot) &: \mathbb{R}^d \rightarrow \mathbb{R}^d \times \mathbb{R}^p \\ \underline{h}_n(\cdot) &: \mathbb{R}^d \rightarrow \mathbb{R}^q \end{aligned}$$

and \underline{u}_n , \underline{v}_n are zero-mean white Gaussian sequences with covariances $Q(n)$, $R(n)$, respectively. The initial state vector \underline{c} is distributed on \mathbb{R}^d according to the density $\underline{J}_0(\cdot)$. If the conditional density of \underline{x}_n , given measurements $\underline{z}_0, \underline{z}_1, \dots, \underline{z}_{n-1}$, is denoted by

$$\underline{J}_n(\underline{y}) d\underline{y} \triangleq \text{prob}(\underline{x}_n \in \underline{y} d\underline{y} | \underline{z}_0, \underline{z}_1, \dots, \underline{z}_{n-1}), \quad (2)$$

then an application of the discrete representation theorem gives the sequential formula for obtaining $\underline{J}_{n+1}(\underline{y})$ from $\underline{J}_n(\underline{y})$ as follows:

$$\underline{J}_{n+1}(\underline{y}) = \frac{C_n}{C_n} \int_{\mathbb{R}^d} T_n(\underline{y}, \underline{z}) \underline{J}_n(\underline{z}) d\underline{z}, \quad (3)$$

where C_n is a normalizing factor, and T_n depends on $\underline{z}_n = \underline{h}_n(\underline{z})$ and $\underline{y} - \underline{f}_n(\underline{z})$ in accordance with the assumed Gaussian density functions². The approximation problem is involved with representing the conditional densities (2) numerically and calculating the d -dimensional integral (3). In many cases of interest the conditional mean is the desired estimate of \underline{x}_n .

3. PROPOSED IMPLEMENTATIONS

Some early attempts to implement approximate optimal nonlinear estimators resulted in the application of power series expansions of the system nonlinearities and/or moment expansions of the conditional density function^{1,5,6}. Such techniques may be referred to by the term "analytical approximations", since the underlying problem is represented by a simpler analytic description. Recently some attention has been turned toward the problem of representing the Bayes integral solution to discrete-time estimation problems by a numerical approximation on a digital computer. Numerical techniques consist basically of two interrelated subproblems - density representation and Bayes-integral computation. For density representation various investigators^{1,7,8} have proposed orthogonal series representations, floating point grid representations^{1,2}, and sums of gaussians¹. The Bayes integral calculation is usually tied very closely to the type of density representation. For example,

orthogonal series representations usually lead to some form of fourier transform calculations for the Bayes integral, either explicitly⁴, or implicitly⁸. Alternatively, the floating point grid representations typically lead to simple quadrature formulas for the Bayes integral. To date, however, the gaussian-sum representation has required the use of analytical approximations such as local linearization, in order to effectively implement the Bayes law computation, which really classifies the technique as a generalized linearized (analytical) technique.

4. AN EXAMPLE PROBLEM

In order to compare the performance of the optimal estimator with that of currently popular approximate estimators we have chosen a simple two-dimensional example which is based on a passive ranging experiment^{1,2}.

$$\begin{aligned}\underline{x}(n) &= F\underline{x}(n-1) + \underline{u}(n-1), & \underline{x}(0) &= \underline{c} \\ \underline{z}(n) &= \underline{h}_n^T \underline{x}(n) + \underline{v}(n),\end{aligned}\quad (4)$$

where

$$\underline{h}_n^T [\underline{x}(n)] = \tan^{-1} \frac{x_2(n) - \sin \beta n}{x_1(n) - \cos \beta n}$$

where the transition matrix F may be taken as the identity in order to simulate random walk, \underline{u} is distributed zero-mean Gaussian with covariance Q , \underline{v} is zero-mean Gaussian with covariance R and β is the angular rotation rate of the sensor. The mathematical model (4) arises in connection with tracking geometries of Figure 1 and Figure 2. In the case of Figure 1, an aircraft sensor S is orbiting about a radarship target T which is undergoing random walk motions from the ocean waves. In this case the initial distribution on \underline{x} , the position of the ship, is taken to be zero-mean Gaussian with identity covariance \underline{I}_0 . The geometry of Figure 1 more closely describes the situation studied previously than that of Figure 2, which illustrates the more important (and obviously more difficult) problem of locating a target T which is separated by a significant distance from the sensor's circular orbit. The problem described by (4)

is in either case one of locating a target by using only bearing sensor information (passive reception). It turns out that the previously studied problem of Figure 1 is considerably easier than originally suspected, and a new experiment comparing all previously proposed methods on the same sample trajectories reveals this fact, as described in the next section.

5. EXPERIMENTAL RESULTS

In order to adequately study the passive receiver problem a number of increasingly difficult cases must be considered. The first case, referred to as the "old problem," is the situation illustrated in Figure 1, with $F = \begin{bmatrix} 0.5 & 0.0 \\ 0.0 & 1.0 \end{bmatrix}$, $\beta = 1$, $Q = \begin{bmatrix} 0.10 & 0.05 \\ 0.05 & 0.10 \end{bmatrix}$, $R = 0.1$, and $\underline{J}_0 \sim N(\underline{0}, \underline{I})$. The second problem, called the "new problem," is identical to the old problem, with the important exceptions that $\underline{J}_0 \sim N\left(\begin{bmatrix} 3 \\ 3 \end{bmatrix}, \underline{I}\right)$, $R = 0.01$, and $F = \underline{I}$. Finally, an example with a multimodal (i.e. nongaussian) initial density, described in Figures 3 and 4, is presented to illustrate what happens if the Gaussian approximation is radically wrong.

5.1 THE OLD PROBLEM

It turns out that the sensor is allowed to have considerable new information with each sample if it orbits about the target at the high rate of $\beta = 1$ radian per sample. Thus it is not surprising to find that it is relatively easy to design an approximate estimator which performs very close to optimum. A straightforward linearized predictor performs miserably, it happens, because of the large fluctuations in the innovations process $\underline{z}_n - \underline{h}^T \hat{\underline{x}}(n)$ as the sensor completes each orbit. In fact, as previously observed^{1,2}, the linearized filter is asymptotically unstable - the mean square error oscillates with ever increasing amplitude. This fact is illustrated in Table 1, where in the second column the Monte Carlo average of the sum squared errors for 100 sample paths is presented for an iterated linearized predictor which is linearized as usual about the current filtered estimate and the filter update is iterated until the sensor linearizations have stabilized. If the

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calculations are continued for more samples in each trajectory then the instability becomes easily evident. The behavior of the linearized predictor is so poor, in fact, that it is worse than the "blind predictor" which is initialized at the mean of the state vector and ignores the data entirely thereafter. The theoretical mean-square error performance of the mean-state predictor is illustrated in column one of Table 1.

If the geometry of the old problem is exploited, it is possible to "fix up" the linearized predictor so that its performance is almost optimal. The fixed linearized predictor is identical to the standard linearized predictor with the addition of logic to calculate the innovations $\underline{z}_n - \underline{h}_n[\hat{\underline{x}}(n)]$ modulo 2π in the range $[-\pi, \pi]$. The sensational improvement in performance of the fixed predictor is illustrated in column three of Table 1.

Finally the numerical techniques of gaussian sums⁵, and floating grids¹, are included in the last two columns of Table 1. If allowances for the fact that the one standard deviation confidence level in a Monte Carlo estimate of N tracks is approximately $2\sigma^4/N$, where σ^2 is the true variance and the densities are assumed Gaussian, then for 100 Monte Carlos the resulting variances are only determined within one standard deviation to ± 17.32 percent of the true σ^2 . Thus, to within the experimental accuracy available in 100 Monte Carlos, the performances illustrated in the last three columns of Table 1 are essentially equivalent. We conclude, therefore, that the problem is not sufficiently difficult to resolve the differences between the various estimation techniques.

5.2 THE NEW PROBLEM

If the geometry of Figure 1 is replaced with that of Figure 2 then it no longer is possible to fix up the performance of the linearized predictor by a simple application of modular arithmetic. Furthermore, the performance of the numerical techniques is decidedly better than the linearized predictor as illustrated in Table 2. The Gaussian sum predictor experienced an unexplained transient error in the first estimate on about the fiftieth

Monte Carlo which was so large that the resulting Monte Carlo estimate is clearly in the transient phase in the table for the first estimate. The recovery is stable, however, so that the remaining values in column three are reliable.

5.3 MULTIMODAL INITIAL CONDITIONS

In the previous two cases the relative success of the linearized-type predictor can be related to the validity of the Gaussian approximation to the conditional density functions. The simplest way to destroy the validity of the Gaussian assumption is to provide a nongaussian initial density function. Consider, for example the detector geometry illustrated in Figure 3. If there were a sequence of known reflecting ionospheric layers above the aircraft observer and we were given an a priori distribution on the transmitter's power, then it is conceivable that we might want to integrate over all elevations to maximize detectability, thus introducing a multimodal range ambiguity as shown in the figure. Probabilistically, the initial condition would be obtained by taking the product of the multimodal range ambiguity density with the bearing ambiguity density. The result might look as in Figure 4, where no particular scale is intended. As soon as the aircraft's detector circuits obtain a reliable detection, the aircraft banks left into a circle of unit radius and activates a high sensitivity receiver which is tuned to one elevation. The resulting problem parameters used for this experiment are $\underline{F} = \begin{bmatrix} 1.001 & 0.000 \\ 0.000 & 1.001 \end{bmatrix}$, $\underline{Q} = \begin{bmatrix} 0.050 & 0.025 \\ 0.025 & 0.050 \end{bmatrix}$, $\underline{r} = .01$, $\underline{s} = .01$, and \underline{j}_0 was chosen with one cross range standard deviation corresponding to 0.25 radian in bearing and a down range standard deviation in each mode of 0.25. If the densities are plotted isometrically as viewed from the direction shown in Figure 4, then some of the more interesting examples from the movie are shown in Figure 5, with sample numbers as given. As seen in Figure 5 the density first peaks up on the wrong range when very little information is available from the measurements, and later makes a dramatic change to the correct mode and stabilizes. This behavior is clearly illustrated in the solid

plot of Figure 6 of the resulting prediction error magnitude. In contrast the dotted curve in Figure 6 shows the performance for the same sample path of the linearized predictor, which is almost identical to the mean state estimate for the problem.

Figure 6 illustrates two interesting phenomena. The linearized approximation can be made useless by only a large uncertainty in initial conditions. On the other hand, the nonlinear estimator, which contains the more detailed information about the conditional densities may still perform satisfactorily. In addition, the optimum system creates a certain amount of optimism for itself initially and calculates a relatively high confidence incorrect estimate. Experience with several such sample paths has resulted in the conclusion that the situation in Figure 6 is just about the worst possible case, that on the average the performance is without much improvement for about half the iterations (0.5 radian of sensor rotations), and then proceeds to converge at about the same rate as the previous examples. The linearized predictor, on the other hand, without a suitable initial condition doesn't ever converge on the average.

6. CONCLUSIONS

The examples have shown the wealth of experimental experience available to the estimation engineer as a result of numerical studies of the problem. It is possible, for example, to assess the validity of given analytical approximations and the value of unused information regarding the structure of the estimation problem. In addition it is possible to compute practical lower bounds on performance for nonlinear estimators in some small dimensional problems, a fact which may have some important engineering significance to designers of sensors, for example, who must know the theoretical limitations on the performance of a given design. It is anticipated that future experiments along these lines will substantiate this conjecture. In the meantime there appears to be many possible application areas to explore.

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Sample	Mean-State Predictor	Iterated Linearized Predictor	"Fixed" Linearized Predictor	Gaussian Sum Predictor	Floating Grid Predictor
1	1.250	4.334	1.468	0.757	0.866
2	1.388	1.678	0.974	0.596	0.674
3	1.447	0.651	0.562	0.457	0.631
4	1.537	0.535	0.550	0.473	0.408
5	1.634	0.828	0.617	0.518	0.403
6	1.734	3.187	0.619	0.531	0.464
7	1.833	4.927	0.540	0.471	0.674
8	1.933	4.912	0.618	0.553	0.454
9	2.033	4.248	0.564	0.578	0.377
10	2.133	3.763	0.619	0.641	0.443
	Unstable	Unstable	Stable	Stable	Stable

Table 1. Monte Carlo Averaged Sum Squared Error
Performance for Predictors - Old Problem

Sample	Mean-State Predictor	Linearized Predictor	Gaussian Sum Predictor	Floating Grid Predictor
1	2.200	1.665	11.436*	0.955
2	2.400	1.928	1.723	1.020
3	2.600	2.187	1.495	1.006
4	2.800	2.229	1.321	1.083
5	3.000	2.190	1.208	1.118
6	3.200	2.145	1.456	1.359
7	3.400	1.830	1.625	1.521
8	3.600	2.105	1.437	1.370
9	3.800	2.136	1.423	1.132
10	4.000	2.349	1.286	1.316
11	4.200	2.370	1.431	1.493
12	4.400	2.114	1.533	1.515
13	4.600	1.883	1.475	1.345
14	4.800	1.752	1.398	1.268
15	5.000	1.752	1.553	1.441
16	5.200	1.850	1.617	1.534
17	5.400	1.627	1.275	1.222
18	5.600	1.561	1.271	1.238
19	5.800	1.620	1.630	1.314
20	6.000	1.688	1.803	1.556

* transient

Table 2. Monte Carlo Averaged Sum-Squared Error
Performance for Predictors - New Problem

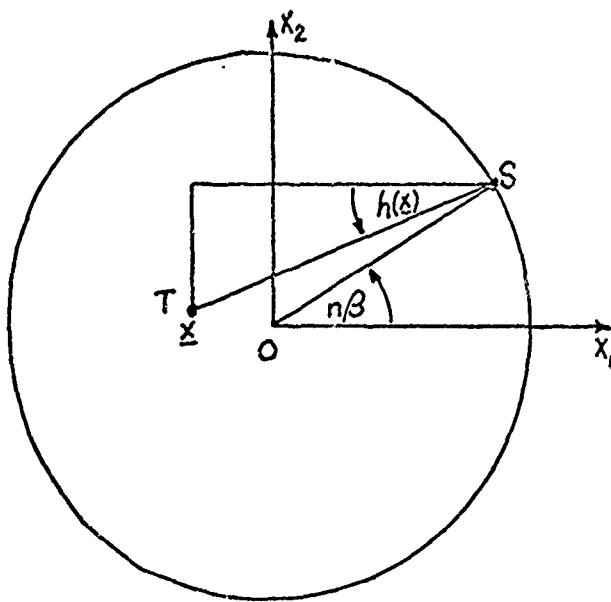


Figure 1. Sensor Geometry - Old Problem

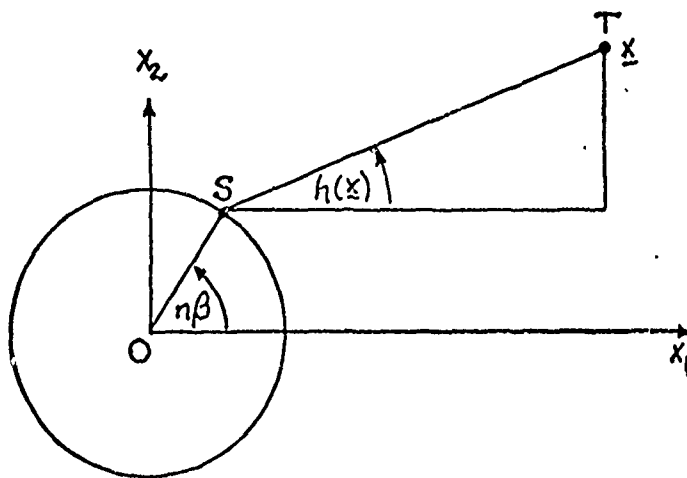


Figure 2. Sensor Geometry - New Problem

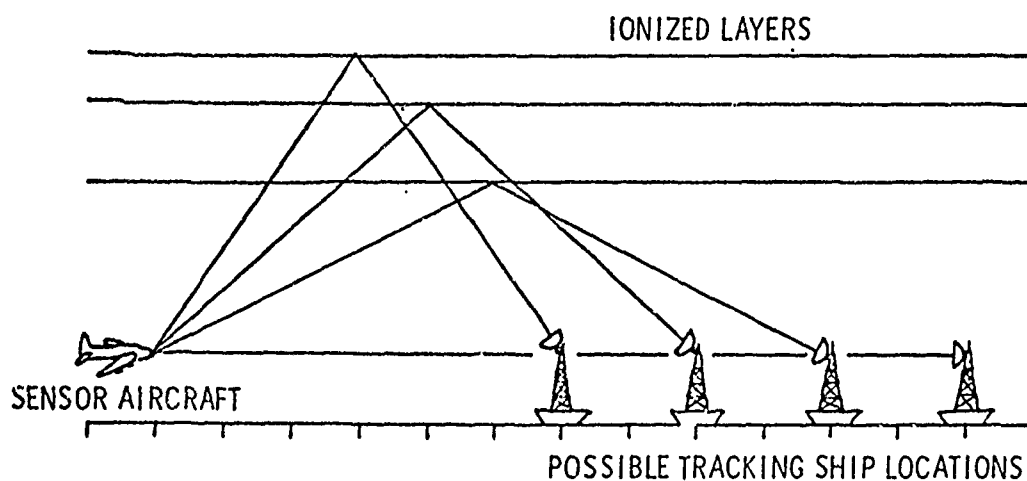


Figure 3. Multimodal Range Ambiguity -
Detector Geometry

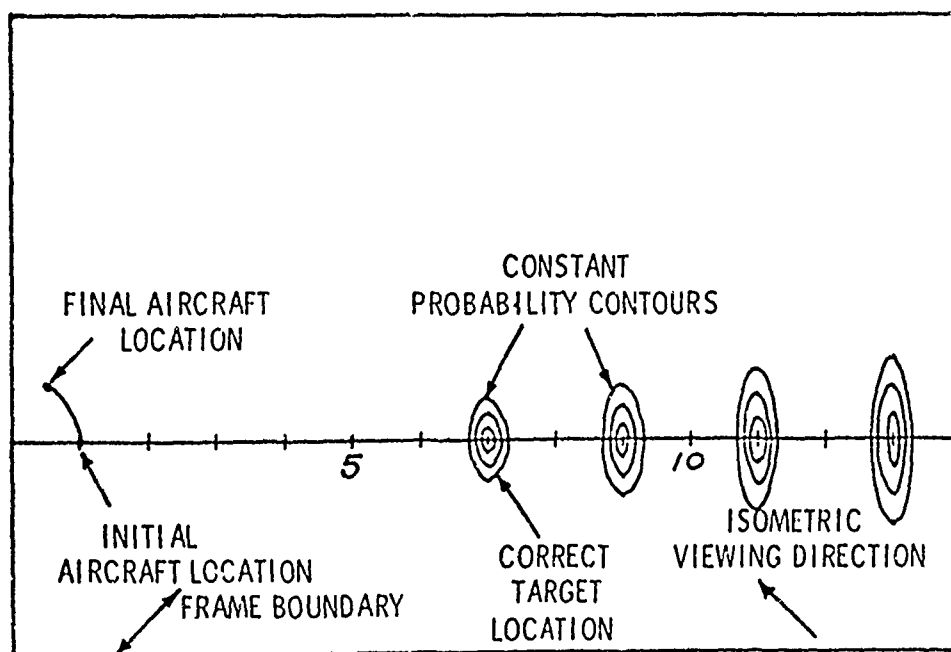


Figure 4. Multimodal Range Ambiguity -
Estimator Geometry

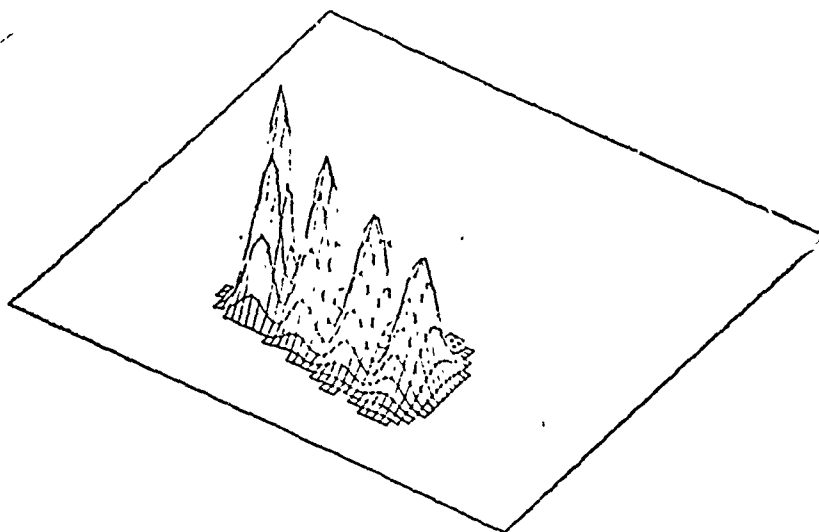
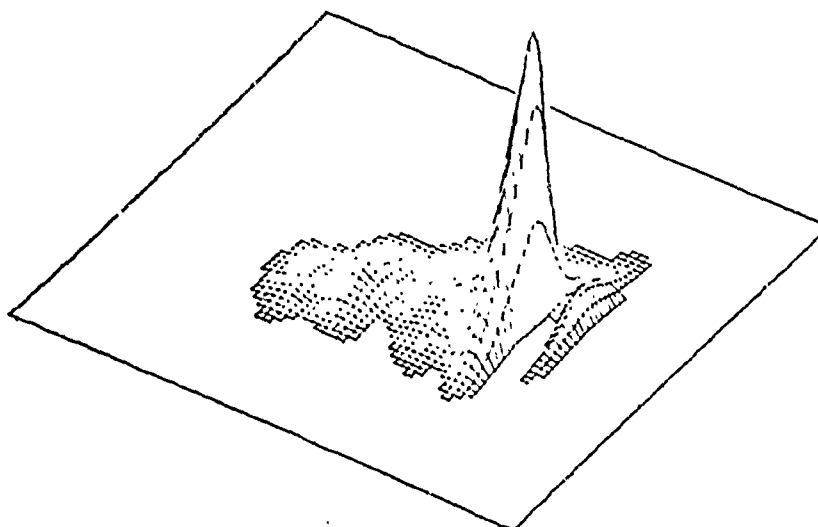
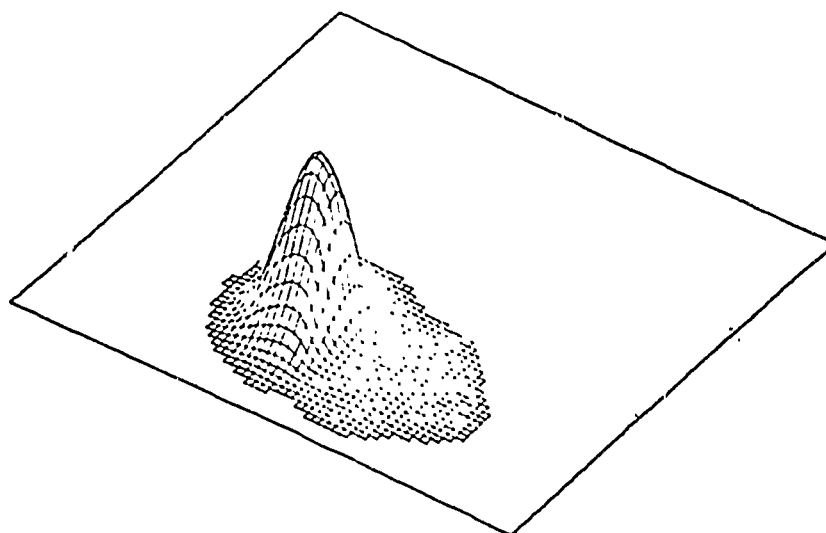
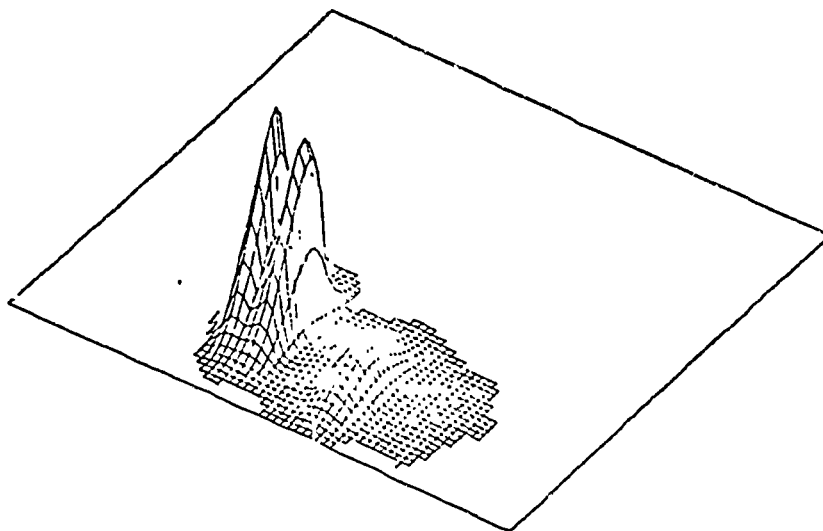


Figure 5. Multimodal Conditional Densities -
Typical Examples:
a. Initial Density b. First Solution
c. Solution Transition d. Final Solution





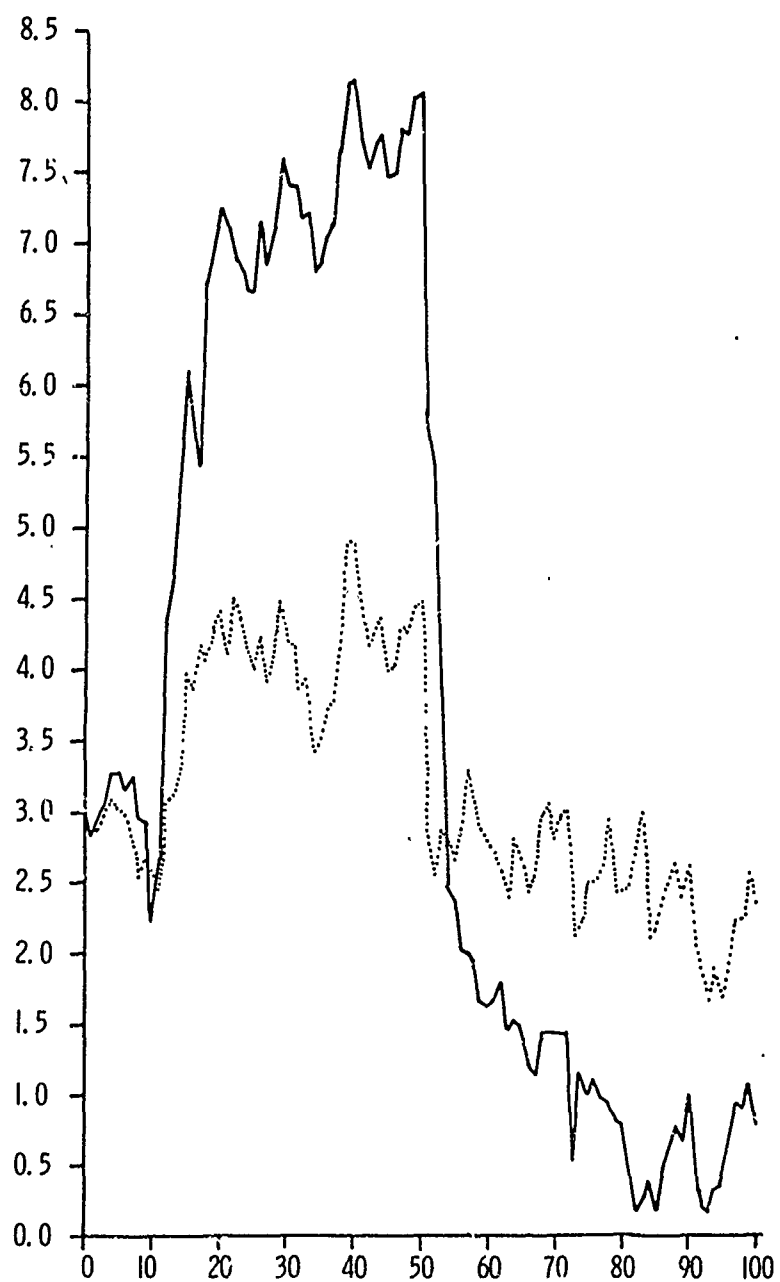


Figure 6. Absolute Error Performance of Optimal and Linearized Predictors for Multimodal Problem